

Between order and disorder: a ‘weak law’ on recent electoral behavior among urban voters?

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Abstract

A new viewpoint on electoral involvement is proposed from the study of the statistics of the proportions of abstentionists, blank and null, and votes according to list of choices, in a large number of national elections in different countries. Considering 11 countries without compulsory voting (Austria, Canada, Czech Republic, France, Germany, Italy, Mexico, Poland, Romania, Spain and Switzerland), a stylized fact emerges for the most populated cities when one computes the entropy associated to the three ratios, which we call the entropy of civic involvement of the electorate. The distribution of this entropy (over all elections and countries) appears to be sharply peaked near a common value. This almost common value is typically shared since the 1970’s by electorates of the most populated municipalities, and this despite the wide disparities between voting systems and types of elections. An even more remarkable stability of this entropy value is observed for the Swiss referendums since the 1880’s.

We suggest that the existence of this hidden regularity, which we propose to coin as a ‘weak law on recent electoral behavior among urban voters’, reveals an emerging collective behavioral norm characteristic of urban citizen voting behavior in modern democracies. Analyzing exceptions to the rule provide insights into the conditions under which this normative behavior can be expected to occur.

Introduction

Each election yields a variable proportion of citizens not taking part in the vote. The proportion of the uninvolved population – either by non-registering, abstaining or voting blank or null – has been much less studied than the vote itself.

Nowadays such behaviors are increasing among the longest-established democracies and their meaning may be changing. Besides passive abstention (due to carelessness or indifference), an active refusal of vote – possibly bearing a political message – is rising among population categories which are usually taking part in the election.

• The modalities of withdrawal [1]

To measure this phenomenon accurately, we first need to define the non-voter turnout. The boundary between voters and non-voters is indeed blurred as several intermediate behaviors exist, such as non-registering or blank vote.

The potential voter population depends on the legal requirements of citizenship, residency and capacity. Registration on the electoral roll does not necessarily imply voting. Moreover, the diversity of enumeration methods from one country to another makes it difficult to compare directly ratios of voters. The main trend consists in comparing abstention to the number of citizens entitled to vote (VEP: Voting Eligible Population). However, in the United States for instance, abstention was calculated until

recently by comparison to the population above the voting age, including foreigners (VAP: Voting Age Population), the corresponding abstention rate often reaching 50%. Another bias stems from the fact that some countries made voting compulsory (namely Belgium, Luxembourg, Greece, and for a time the Netherlands, Austria and Italy). Without compulsory voting, a declining voter turnout is observed since the 1980s in established democracies.

Moreover, the meaning of blank and null vote is not obvious. They could be considered at first sight as equivalent to abstention or non-registering, since they seem to translate an absence of choice. This hypothesis would be in agreement with the systematic reviews of the minutes of polling stations for instance.

Abstention has been primarily considered to be a micro-level phenomenon. But is it really? Several studies have proven that socio-economic characteristics such as gender [2, 3], age [4], education [5, 6] and ethnicity [7] have an influence on electoral non-participation. To what extent does living in a community with low level of electoral involvement influence a voter?

• The political and institutional context of the election

The comparative database collected by the Institute for Democracy and Electoral Assistance (IDEA [8]) gathered data from elections in 171 countries from 1945 to 1999. It shows that participation rates are slightly higher in countries that have adopted a system of proportional representation, offering a larger choice to voters than those which have a majority or mixed systems. The highest turnout recorded (over 83% observed in both Malta and Ireland) corresponds to the system of ‘single transferable vote’ which gives the voter a large liberty margin.¹

The nature of the election may be important too, depending on the context. In France for example, as the president has a lot of power, the participation rate of the presidential election is especially high when compared to the parliamentary election.

• Abstention and Blank and null votes

The reason why analysis of political sciences are paying little attention to blank and null votes is mostly based on the fact that these ballots are representing a very small number. Typically, these votes are aggregated within a single category, Blank and null votes, in some countries simply called Null (or Invalid) votes. Multitudinous studies have demonstrated from the 1950s on that null ballots were subdivided at random, according to the law of large number and distributed haphazardly for a given manner of voting [9]. The analysis of each voting office is still confirming that. However, the blank votes are more sensible to the conjuncture of consultation and are taking, with regard to abstention, a more complex signification.

Voters casting a blank vote are having motivations closer to voters abstaining for political reasons. This “civic abstention”, as Alain Lancelot called it, translates a particular attitude regarding the voting procedure [9, 10]. Statistical analysis of abstention, blank and invalid votes show a negative correlation, often quite important, between these two ways to not take part in the election. In France it has often been observed, the more abstention is important, the more voters are living in most populated municipalities. Converse argument, the more they are living in rural area, the less abstention is pronounced. On the other hand if blank and null ballots are less numerous in large agglomerations, their number is showing an upward trend in smaller municipalities. The “civic abstention” is playing an important part there. This correlation between abstention and blank and null ballots shows a tendency to complexify. The political attitude of “withdrawal” or political “offside” is less easy to analyze. The urbanization has led to important changes in lifestyle and therefore in the voting behavior. We will analyze the situation in some countries, for all votes we have been able to obtain data, and hence try to better understand this interrelation between abstention, blank and null ballots and the expression of the vote, if existing.

• Stylized facts

In this paper we consider together the three values: abstention, blank/null votes and total valid voters. The focus will be on the identification of statistical regularities, in the spirit of recent statistical physics

¹This system, called Hare system of voting, is a variant of proportional representation where the voters rank the candidates according to their preferences.

analysis of elections data – see e.g. [11–23].

In the present work, by analyzing a large number of elections in 11 different countries without compulsory voting, we point out that they share a common feature. Introducing a measure of civic involvement of electorate, we show that this quantity exhibits a sharply peaked distribution around a common value in highly populated municipalities in recent time. Moreover we suggest that this common stylized fact, that we denote ‘weak law on recent electoral behavior among urban voters’, reveals an emerging collective behavioral norm, typical of citizen voting behavior in modern democracies.

The paper is organized as follows. First we describe the dataset used in this study, at three different scales (at the municipality scale, at larger scale but for older times, and at the polling station level when it is possible). Then, we introduce and discuss what we call the involvement entropy. We then analysis electoral data according to this measure, and give signs of existence of a possible norm revealed by a common-value of this measure. Supporting Information (SI) gives more details when it is necessary.

Results

Dataset

In this paper we analyze electoral data at three different scales. (1) Data aggregated at the municipality scale. By this way, we study phenomena with respect to the population size of municipalities. The 76 elections studied in this paper at municipality level are mostly recent, after 1990, and are taken from 11 different countries (Austria, Canada, Czech Republic, France, Germany, Italy, Mexico, Poland, Romania, Spain and Switzerland). (2) Electoral data aggregated at large scale, e.g. national, provincial, etc. Here, we focus the analysis on time evolution. Countries studied for their historical aspects are those which are studied at the municipality scale. The study begins at the earliest year as possible, i.e. at the beginning of so-called democratic regimes, after World War II, and even earlier for some cases (e.g. 1884 for the ≈ 530 Swiss referendums). (3) Electoral data aggregated at the polling station level. Polling stations over the 100 most populated municipalities are analyzed, whenever it is possible to do so (i.e. for Canada, France, Mexico, Poland and Romania). Some intra-towns phenomena are investigated by this way.

Some elections are studied as a function of the number N of registered voters by municipality. This is the case when the following conditions are valid: (1) elections in a democratic country with no compulsory voting, and no duty against people who do not vote; (2) the number of registered voters by municipality is well established ²; (3) available data provide for each municipality, at least, the number of registered voters, the number of votes or the turnout rate, and the number of valid votes. We note that all countries for which we have the data at the municipality scale have more than 2000 municipalities, which allows us to make statistical analysis. Moreover, all elections studied here are national ones, except for *Land*

²In particular this excludes from our study both the U.S.A. and England.

At	13	1945	Ca*	5	1945	CH	3	1884	Cz	1	1990	Fr*	20	1946	Ge	7	1949
It	4	1946	Mx*	4	1991	Pl*	11	1990	Ro*	4	1990	Sp	4	1976			

Table 1. Countries where elections are analyzed in this paper (first column). Number of elections studied at the municipality scale (second column), and the date from which they are studied at national or provincial scale (third column) – even if it is before the end of the compulsory voting in Austria and in Italy. Star indicates that electoral data are also known at polling station level. Number of municipalities per country: ≈ 2400 in Austria (At); ≈ 7700 in Canada (Ca); ≈ 2700 in Switzerland (CH); ≈ 6400 in Czech Republic (Cz); ≈ 36000 in Metropolitan France (Fr); ≈ 12000 in Germany (Ge); ≈ 8100 in Italy (It); ≈ 2400 in Mexico (Mx); ≈ 2500 in Poland (Pl); ≈ 3200 in Romania (Ro); ≈ 8100 in Spain (Sp). See in the SI, Section A for more details.

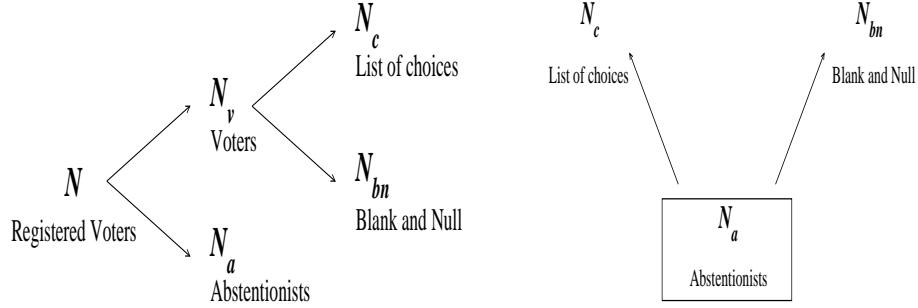


Figure 1

Parliament elections in Germany. Lastly, the choice of the studied elections is not rooted on a plan but simply on the availability of electoral data.

Among these 76 elections, 31 of them are also analyzed at polling station level in the 100 most populated town: 5 from Canada (≈ 25000 polling stations), 13 from France (≈ 7000 polling stations), 4 from Mexico (≈ 55000 ballot boxes), 11 from Poland (≈ 8000 polling stations), and 4 from Romania (≈ 6000 polling stations). Tab.1 summarizes the set of elections studied in this paper, and more details on these data are given in the SI, Section A.

Abstentions, valid votes and blank or null votes

Let us describe the citizen classification here retained to characterize the electoral mobilization of registered voters. For each given election and each specific scale (a municipality, a province, a country, etc.) we distinguish: (1) the total number N of registered Voters; (2) the number N_a of Abstentionists, the persons who do not take part to the election; (3) the number N_v of voters, among which (4) N_{bn} Blank and Null Votes ³ and (5) N_c Votes in favor of candidates or electoral list of choices, also sometimes called Valid Votes (see Fig. 1). Obviously $N_v = N_c + N_{bn}$ and $N = N_a + N_c + N_{bn}$. Note that in Italy, Spain, and Switzerland, electoral data distinguish between Null Votes, N_n , and Blank Votes, N_b . Moreover, only in Spain, “*Votos Válidos*” means $N_v - N_n$, that differs from other countries where “Valid Votes” means $N_v - N_n - N_b$. In this paper, we consider for all countries that Valid Votes are defined as $N_c = N_v - N_{bn}$. See in the SI, Section F for more discussion about countries where Blank Votes and Null Votes are distinguished between each other.

As discussed in the following, we characterize the civic involvement of registered voters by the choice between the three possible sates, Abstention, Blank or Null Vote and Valid Vote. The civic involvement of electors is then here measured through the set of the three ratios $\{p_a, p_c, p_{bn}\}$, defined by

$$p_a = \frac{N_a}{N}, \quad p_c = \frac{N_c}{N}, \quad p_{bn} = \frac{N_{bn}}{N}, \quad (1)$$

with $p_a + p_c + p_{bn} = 1$. Each election can then be represented by a point in the simplex $p_a + p_c + p_{bn} = 1$, as illustrated on Fig. 2. Since the number of Blank and Null is typically small, clearly most points lie near the edge $p_{bn} = 0$. However, here we do not want to neglect this component (see below for a deeper discussion). A second basic observation is that there is a wide dispersion along the axis $p_a - p_c$.

Previous work [22] has revealed strong regularities in the fluctuations around the mean of $p_v \equiv N_v/N$, more exactly of the logarithmic turnout rate $\tau \equiv \log \frac{p_v}{1-p_v}$, when looking at its distribution over municipalities, and particularly over the most populated municipalities. Similarly, a logarithmic three

³Some countries, like Canada and Poland, aggregate Blank and Null votes in an only one term called as Null votes, or Invalid votes, or Spoilt votes.

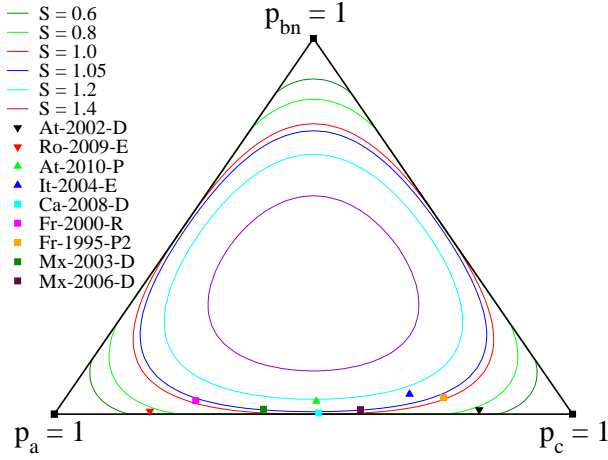


Figure 2. Simplex $p_a + p_c + p_{bn} = 1$, in which any given election, for the most populated municipalities, can be represented by a point, as illustrated by the symbols corresponding to particular elections of our data set (see the SI, Section A for details). The continuous curves are lines of constant involvement entropy value, drawn for values ranging from $S = 0.6$ to $S = 1.4$. See text for At-2002-D, Ro-2009-E, At-2010-P and It-2004-E. For Ca-2008-D: $p_a \simeq 0.49$, $p_c \simeq 0.51$, $p_{bn} \simeq 0.003$ and $S \simeq 1.02$; for Fr-2000-R: $p_a \simeq 0.71$, $p_c \simeq 0.25$, $p_{bn} \simeq 0.036$ and $S \simeq 1.02$; for Fr-1995-P2: $p_a \simeq 0.23$, $p_c \simeq 0.73$, $p_{bn} \simeq 0.044$ and $S \simeq 1.01$; for Mx-2003-D: $p_a \simeq 0.59$, $p_c \simeq 0.40$, $p_{bn} \simeq 0.013$ and $S \simeq 1.04$; and for Mx-2006-D: $p_a \simeq 0.40$, $p_c \simeq 0.58$, $p_{bn} \simeq 0.012$ and $S \simeq 1.04$.

choices value can be defined, for which the same type of regularities can be observed when considering polling stations within municipalities (see the SI, Section D). Moreover, this analysis of fluctuations suggests that individual behavior is not well explained by a sequential binary choice (to vote or not, then to cast a valid vote or not). This, with the following analysis, justify to consider together the three quantities p_a , p_c and p_{bn} . Hence the *electoral involvement* should be viewed through the three possibilities available to the voters: abstention, blank/null votes and votes according to the list of choices.

In addition, analysis of fluctuations (see the SI, Section D) tells nothing about the *mean* value itself. In this paper, we will precisely be interested in the properties of mean values.

The involvement entropy

We introduce a variable whose value, as we will argue, is appropriate for characterizing the mean civic involvement of the electorate. Viewing the three ratios $\{p_a, p_c, p_{bn}\}$ as probabilities, it is interesting to associate to each election, instead of these three numbers, a single scalar characterizing the probability distribution itself. One natural quantity associated to a probability distribution is the entropy, S , defined by

$$S(p_a, p_c, p_{bn}) = -p_a \log(p_a) - p_c \log(p_c) - p_{bn} \log(p_{bn}). \quad (2)$$

Here, and throughout this paper, \log means base-two logarithm ($\log(2) = 1$, and the entropy is said to be in units of bits).

Within the framework of Information Theory, where it is called the Shannon entropy, this quantity can be understood as a measure of missing information, or of average surprise, associated to the studied random process [24]. In the context of Statistical Physics, it is the Boltzmann-Gibbs entropy measuring the degree of ‘disorder’ of the system under consideration [25]. In the present context, we will refer to S as the entropy of civic involvement, or “involvement entropy”, and consider it as a measure of disorder vs. order in the civic involvement at a collective level. Indeed, it is a ‘macroscopic’ or collective measure about the civic involvement of an electorate, and not the measure of the civic mobilization of individual citizen – i.e., we do not claim that it corresponds to the behavior of a representative citizen. It can be measured at any scale of aggregate data, e.g. for a municipality, a province, or a whole country. For instance, the involvement entropy of a municipality, S , is given by Eq. (2) where the three ratios

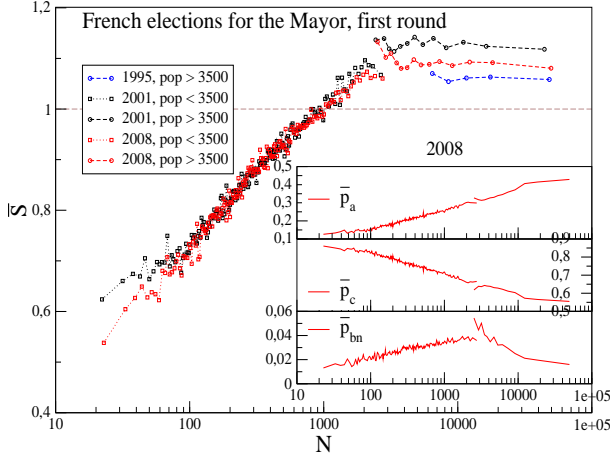


Figure 3. Average values \bar{S} of the involvement entropy of municipalities, S , as a function of the number of registered voters N , for the first round of Mayor elections in France. There are two kinds of voting rules, which depend on the population-size more or less than 3500 inhabitants (see text).

Inset shows average values of p_a , p_c and p_{bn} as a function of N for the 2008 *municipales* elections (which lead for high population municipalities to a plateau of S despite variations in p_a , p_c and p_{bn}). For each N , average values, \bar{S} , \bar{p}_a , \bar{p}_c , \bar{p}_{bn} , are evaluated over ≈ 200 municipalities of size $\approx N$.

$\{p_a, p_c, p_{bn}\}$ are the ratios of, respectively, the number abstentionists, N_a , valid votes, N_c , and blank and null votes N_{bn} , over the total number N of registered voters in the considered municipality.

Let us explain more what we mean by ‘order/disorder’, and how this is reflected by the entropy value. We consider that a civic involvement shows an ‘ordered’ state if one of the three ratios is very close to one (hence the two others very small). A ‘disordered’ state corresponds to having all three ratios of similar values. Within this viewpoint, no particular role or importance is assigned to any one of the three possible cases, abstention, blank/null, valid vote. The involvement entropy S , a positive or null quantity, provides a well defined way to quantify the degree of disorder: the larger the entropy, the larger the disorder. The maximum order is obtained when one of the ratios is equal to unity (and then the two others are equal to zero), in which case $S = 0$. In contrast, the maximum disorder corresponds to an equipartition of these 3 ratios, that is $p_c = p_a = p_{bn} = 1/3$, in which case the entropy takes its maximal possible value, $S = \log(3) \simeq 1.58$.

As an illustration, consider the elections for the Mayor in the French municipalities. It is well known (at least in France) that participation to elections in small municipalities is typically larger than in large cities, for social reasons – for instance, in small municipalities where everyone knows every one else, not going to the polling station will become common knowledge. Such social enforcement of the civic involvement might be at the root of an increase of the number of abstentionists with population size: the ratio p_a of abstentionists is typically very low for small municipalities, and increases with the municipality size, N . One then expects an increase of the involvement entropy with municipality-size: this is indeed what we observe for the elections for the 2001 and 2008 first round (elections for which we have the data for all the municipalities), as illustrated on Fig. 3. We can say that the electorate is very “ordered” (in terms of its civic involvement) for low municipality-size, and gets more “disordered” with increasing N . This involvement entropy increase is observed until a threshold population size value, at which the electoral rule changes: the citizen has a larger number of possible voting choices in municipalities with a number of inhabitants smaller than 3500, than in more populated municipalities⁴. Remarkably, above this critical size, the involvement entropy becomes essentially independent of the population size: one has a plateau, at S slightly above 1, despite variations in p_a , p_c and p_{bn} . As we will see throughout this paper, this particular value of involvement entropy, $S \approx 1$, shows up as a typical value in modern elections for most populated cities.

Let us give other illustrations. A great order of the electorate is provided by: (1) the population of registered voters is highly polarized: there is an important difference between p_a and p_c ($p_a \ll p_c$ or

⁴It is allowed for citizens living in municipalities with less than 3500 inhabitants, to combine candidates from different opposite lists, or to add new names from citizens who are not officially candidates.

$p_a \gg p_c$); and (2) blank and/or null votes are very few, that is p_{bn} is very small. Such cases of small entropies are, e.g., the 2002 Austrian Chamber of Deputies election for which $p_a \simeq 0.17$, $p_c \simeq 0.81$, $p_{bn} \simeq 0.011$ and $S \simeq 0.73$; the 2009 European Parliament election in Romania, with $p_a \simeq 0.81$, $p_c \simeq 0.18$, $p_{bn} \simeq 0.008$ and $S \simeq 0.73$. Conversely, a great disorder of the electorate results from: (1) the population of registered voters is not very polarized, that is p_a and p_c are not very different; and (2) blank and/or null votes are relatively important, that is p_{bn} is not too small. For instance, the 2010 Austrian Presidential election has $p_a \simeq 0.48$, $p_c \simeq 0.49$, $p_{bn} \simeq 0.034$ and $S \simeq 1.16$; and the 2006 European Parliament election in Italy has $p_a \simeq 0.29$, $p_c \simeq 0.66$, $p_{bn} \simeq 0.053$ and $S \simeq 1.11$. Note that these values come from great town values (see the SI, Tab. S1), whereas S is more spread out in small municipalities (see Fig. 4). Finally, one finds that the involvement entropy S has a value frequently very near 1.0. For example, the 2008 Canadian Chamber of deputies election, the 2000 French referendum, the 1995 French second round Presidential, and the 2003 and 2006 Mexican Chamber of deputies elections (see Fig. 2 and the SI, Tab. S1). In all these examples, despite an important diversity in p_a values, S lies within 1.01 and 1.04, showing that the electorate polarization is somewhat halfway between order and disorder. Note that $S = 1$ is the entropy associated to the tossing of a fair coin. In the present context, it would be exactly obtained for elections with $p_a = p_c = 50\%$ and $p_{bn} = 0$.

Data Analysis

We have computed the involvement entropy S for all the elections of our data set, at different scales. First we find that, most often, it depends on the municipality-size N . To analyze this size dependency, we spread out municipalities data over samples with respect to the municipality population-size. In each sample, municipalities have roughly the same number of registered voters. The number of municipalities per sample is of order 100, except for France in which case this number is 200 (because France has much more municipalities than the other countries studied in this paper). We denote⁵ by \overline{S} the average over all municipalities inside a sample of the involvement entropy S . This average \overline{S} is plotted in Fig. 4 as a function of the number of registered voters, N .

Let us give the 1995 French second round Presidential election (Fr-1995-P2) as an example. A relatively ordered civic electorate involvement is observed for the smallest population-size municipalities, with $\overline{S} \simeq 0.7$. The mean involvement entropy then increases with municipality size, for sizes up to $N \sim 10000$. For the most populated municipalities, that is above this threshold value in population-size, a saturation occurs: the (average) civic disorder of the electorate becomes independent of municipality-size, with $\overline{S} \approx 1$.

Next we consider the time evolution of the involvement entropy at a large scale (country, province, *canton*, etc.). When the scale of aggregate data is lower than the national one, each value of the involvement entropy for one election is equal to a weighted (by population-size) mean value of involvement entropies at lower scale (province, *canton*, etc.). (See the SI, Section A and Tab. S2 for more details.) In the SI, Fig. S1, plots the involvement entropy of each election at large scale, for each country over all elections (according to its nature) as a function of time, and Fig. S2 shows how p_a and p_{bn} evolve in time for Chamber of Deputies election in each country. Nevertheless a rapid evolution in time of S can be seen in Fig. 5, where elections in each country are divided into two groups (with roughly the same size each): the older ones and the most recent ones. This figure shows histograms of the involvement entropy at large scale (and also for p_a , p_c and p_{bn}) of 321 elections, according to their relative position in time for each country. From the relative older elections to the more recent ones, a significant peak appears for $S \approx 1$, that we called an halfway between order and disorder. This point mainly occurs in parallel with the significant decrease in time of high ordered elections (in the civic involvement point of view). In other words, nowadays there are few elections with a small civic involvement entropy, S (say

⁵In this paper, \overline{X} means the average value of the considered value, X , over all municipalities (around 100, or 200 for France) in a given sample where municipalities have roughly the same number of registered voters, N ; e.g. \overline{S} , $\overline{p_a}$, $\overline{p_{bn}}$, etc.

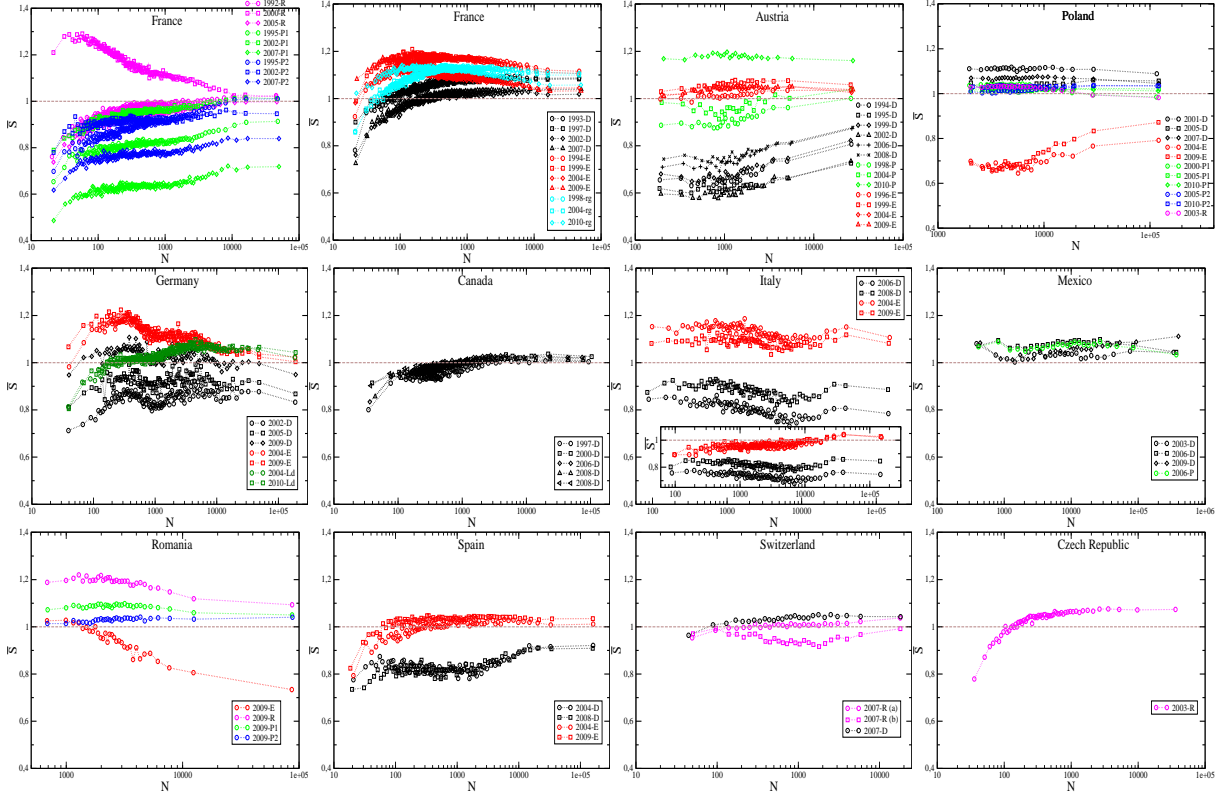


Figure 4. Mean values \bar{S} of the involvement entropy for municipalities, as a function of the number of registered voters N . Each point results from an average over a sample of ≈ 100 (200 for France) municipalities of size $\approx N$. Italian graph inset shows a variant of S where Blank Votes are grouped with Valid Votes (see Section F for a deeper discussion). See the SI, Section A and Tab. S1, for more details on the data.

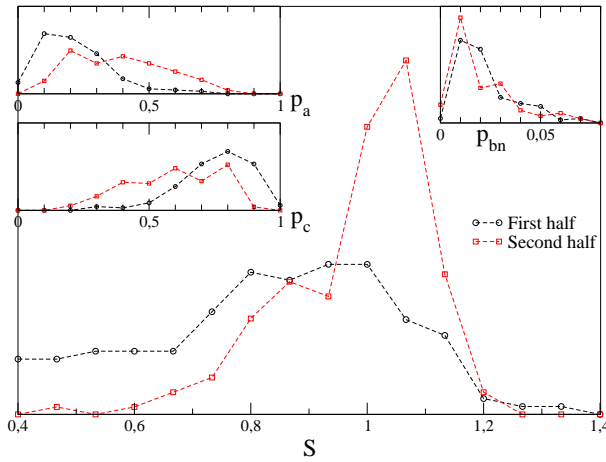


Figure 5. Evolution in time of involvement entropy, S , at large scale (national, provincial, etc.) of 321 elections (see the SI, Section A and Tab. S2, for more details), apart from Swiss referendums. For each country, electoral results are equally divided into two groups: those which occurred at the first period in time and at the second one. Histograms of S (and p_a , p_c and p_{bn} in the insets) show the involvement entropy of the first and second group over all countries. See The SI, Fig. S1 which plots for each country the whole of elections, and also Fig. S4 for scatter plots (p_a , p_{bn}) of these elections, but at national aggregate scale.

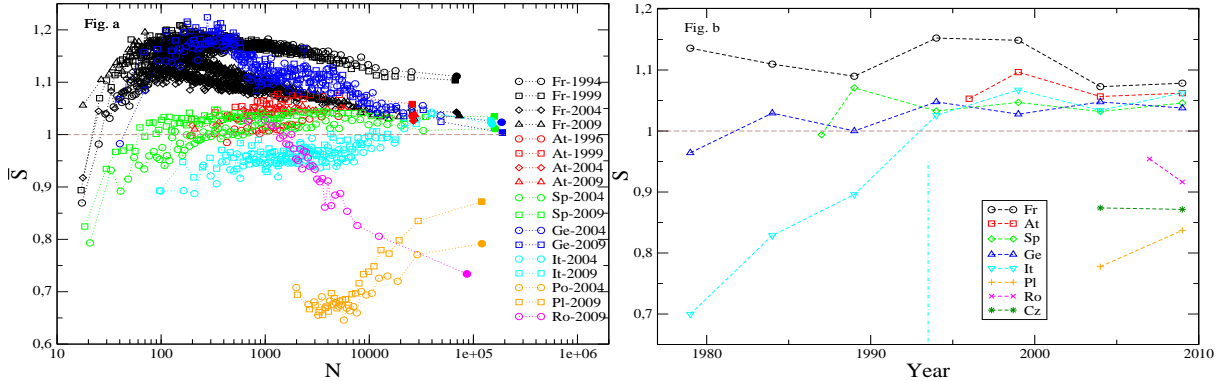


Figure 6. Mean involvement entropy for European Parliament elections. Fig. 6-a (left panel): same elections as those shown in Fig. 4; here for all countries, including France, averages are over ~ 100 municipalities. Fig. 6-b (right panel): same elections as those shown in the SI, Fig. S1; the vertical dashed line indicates the year of the abolishment of compulsory voting in Italy. Here, Italian Blank Votes, N_b , (but not Null Votes, N_n) are grouped with votes in favor of lists of candidates (see the SI, Section F for more discussion).

e.g. $S \lesssim 0.8$), but there are a lot of elections with $S \approx 1$.

Finally, Fig. 6 shows, for all the European Parliament Elections, how the involvement entropy of municipalities depends on population-size (like in Fig. 4), and the time evolution at the national or provincial scale (like in the SI, Fig. S1).

The common occurrence of $S \approx 1$

What the common occurrence is

As already said, Fig. 4 shows the remarkable fact that, for each studied country, in modern elections the involvement entropy of highly populated municipalities is very frequently roughly equal to 1. This common value, $S \approx 1$, for high population-size municipalities is particularly striking when one looks at European Parliament Elections (see Fig. 6-a). See also Table 2 for a rapid overview and basic statistics per country about involvement entropies and population size of the ≈ 100 most populated municipalities. There are however noticeable exceptions, notably the Italian case on which we will come back later (Section Discussion). In any case, we have now to better specify what we mean by $S \approx 1$ and show more quantitatively in which way it is a common properties of modern elections. This is done by gathering data over all elections after 2000 (after 2000 in order to take into account evolution in time of the involvement entropy as stressed by Fig. 5 and Tab. 2). Fig. 7-d plots the resulting histograms of the involvement entropy restricted to ≈ 100 most populated municipalities, for different countries or ensemble of countries. Moreover, Fig. 8 shows respectively the minimal length interval of S , p_a , p_c and p_{bn} which contain 50% of events (those plotted in Fig. 7-d). These two figures show a common sharp peak at a value of S close to 1. The involvement entropy appears to be mainly in the range $0.98 \lesssim S \lesssim 1.08$, which can be taken as the definition of $S \approx 1$ in this paper. Note that this definition is applied to the most populated municipalities. At large scale, the involvement entropy depends on the way that data are aggregated (at national, province, etc. scale), and it is a little bit greater than \bar{S} for the most populated municipalities. Nevertheless the involvement entropy measure at large scale approximately reflects how the most populated municipalities do, because an important ratio of population live in the ≈ 100 most populated municipalities (as seen in Tab. 2).

It is important to stress that the common occurrence $S \approx 1$ appears (1) as a property of high

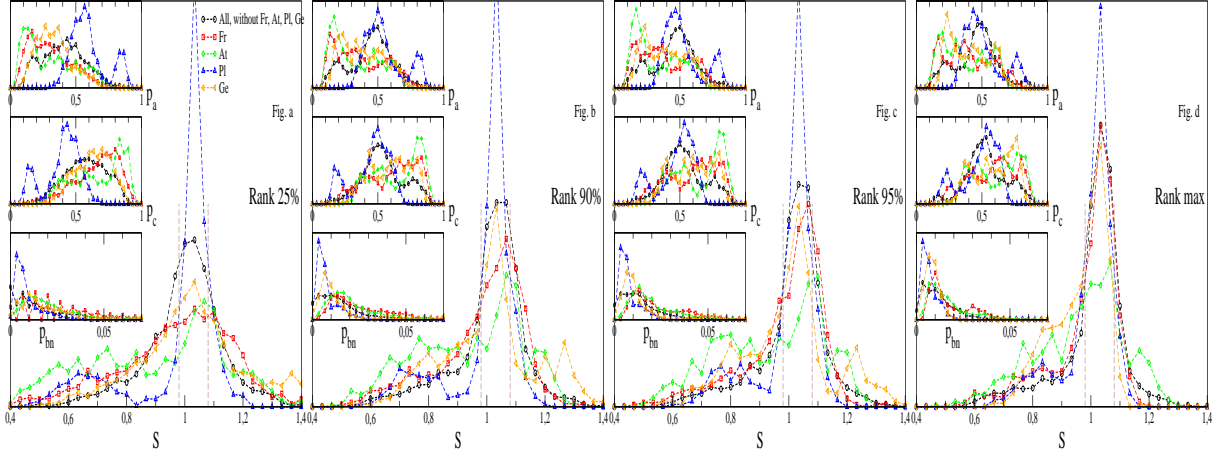


Figure 7. Histograms of involvement entropy, S , with respect to the relative municipality-size bin over all analyzed since 2000. There are 12 French elections, 7 Austrian elections, 11 Polish elections, 7 German elections and 24 for others countries (included in one curve, with no more than 4 elections per country). Municipalities of each country are divided into bins (of ≈ 100 municipalities) with respect to their municipality-size (see e.g. Fig. 4). For instance, ‘Rank 25%’ (Fig. 7-a) means the bin whose population-size rank is the twenty-fifth per cent with regard to the sample of the most populated municipalities (Fig. 7-d) of the considered country. Insets: histograms of corresponding p_a , p_c and p_{bn} . $S = 0.98$ and $S = 1.08$ are plotted in dashed lines and all the scales axis are similar from one plot to another one.

Ctry	date	n_{el}	N_{min}	\overline{N}	N_{bin}/N_{Ctry}	$\overline{S} < 0.98$	$\overline{S} \in [0.98, 1.08]$	$1.08 < \overline{S}$
$t < 2000$								
Fr		8	32000	69000	18%	1	2	5
At		6	7000	26000	44%	3	3	0
Ca		1	30000	94000	48%	0	1	0
$t \geq 2000$								
Fr		12	33000	70000	17%	3	8	1
At		7	7000	27000	43%	3	3	1
Pl		11	39000	120000	39%	2	8	1
Ge		7	68000	190000	30%	3	4	0
Ca		4	53000	83000	38%	0	4	0
It		4	48000	150000	31%	2	0	2
Sp		4	48000	160000	47%	2	2	0
Mx		4	130000	370000	53%	0	3	1
Ro		4	20000	87000	47%	1	2	1
CH		3	7500	18000	37%	0	3	0
Cz		1	14000	36000	43%	0	1	0

Table 2. Basic information about the bin of the ≈ 100 most populated municipalities per country (Ctry). n_{el} means the number of elections analyzed. The municipality of this bin with the lowest number of registered voters is written as N_{min} ; the average value of N over these municipalities, as \overline{N} ; and the ratio of registered voters which belongs to this bin over those in the whole country, as N_{bin}/N_{Ctry} . \overline{S} is classified according to values 0.98 and 1.08.

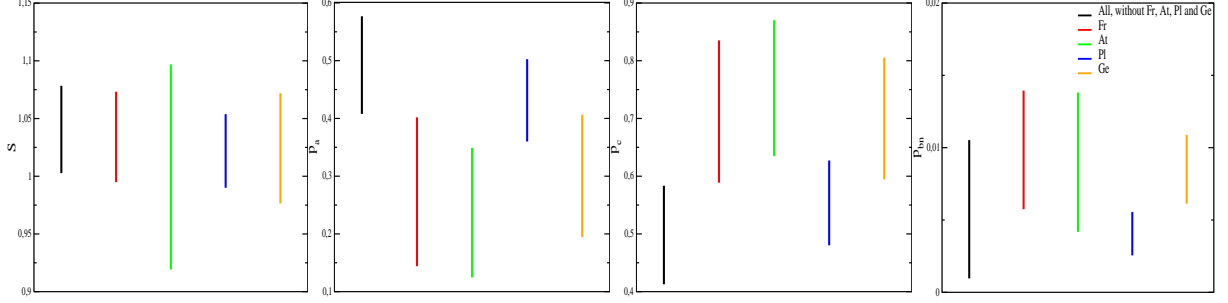


Figure 8. Minimal intervals containing 50% of events, of the involvement entropy S and ratios p_a , p_c and p_{bn} of the 100 most populated municipalities over all elections since 2000. See Fig. 7-d for the related histograms.

populated municipalities, (2) and also in a recent time. See Fig. 5, or the SI, Fig. S1 as an indication of the latter point. For the first point, Fig. 7 shows the histograms of the entropy for different municipality sizes. Compared with histograms of the most populated municipalities (Fig. 7-d), histograms of lower municipality-size appear: (1) much less peaked (apart from Polish elections), and (2) not peaked at the same common-value. Moreover, it is only for the larger sizes that all the histograms become very similar, suggesting the convergence to a universal histogram at large sizes. Let us bear in mind (cf. Tab. 2) that the sample of the ≈ 100 most populated municipalities in Austria is, on average, much less populated than the ones of the four other countries or ensemble or countries.⁶ In other words, the Austrian sample of the ≈ 100 most populated municipalities is not so comparable to the four other ones, and does not accurately reflect a typical behavior in large populated municipalities (especially since the civic involvement can significantly depend on the population size as it is shown in Fig. 4). Lastly, the choice of the number (here 100) of most populated municipalities is only for statistical convenience and does not affect the results (see e.g. the SI, Fig. S3, which is similar to the Fig. 7-d, but for the sample of 50 or 200 most populated municipalities).

Now, let us better quantify this sharp and common peak for the most populated municipalities. First, Fig. 9-a plots the smallest distance ($S_{sup} - S_{min}$), such that 50% of events are included into the set $[S_{inf}, S_{sup}]$, with respect to the relative municipality size. This confirms that (apart from Polish elections) distributions of S get more peaked when the population size increase, and specifically for the most populated municipalities.⁷ Moreover (apart from the Austrian elections) the minimal distance ($S_{sup} - S_{min}$) appears to converge to a common value, this only for the most populated sample (see also Fig. 8 for S_{inf} and S_{sup} for this latter sample). Next, in order to quantify the common peak phenomenon, we calculate the overlap between distributions of S for municipalities as a function of the relative population size (see Fig. 9-b). The overlap between n distributions of S , with probability density functions (pdf) $f_i(S)$, $i = 1, 2, \dots, n$, is defined as $\mathcal{O}_n = \int \min[f_1(S), f_2(S), \dots, f_n(S)] dS$. Fig. 9-b shows an increasing overlap between distributions when the population size increases, and specifically for the most populated municipalities. This confirms that the distributions of S get more and more similar as the relative municipality-size increases, with (sharp) peaks becoming identical for the most populated municipalities.

⁶Taking into account the ≈ 50 Austrian municipalities per sample provides, for the most populated sample, an histogram of S much centered on $S \approx 1$ than the one of ≈ 100 municipalities (see the SI, Fig. S3).

⁷The same features also appear by considering minimal distances which contain 25% or 10% of events. This is in agreement of the robustness of this trivial method.

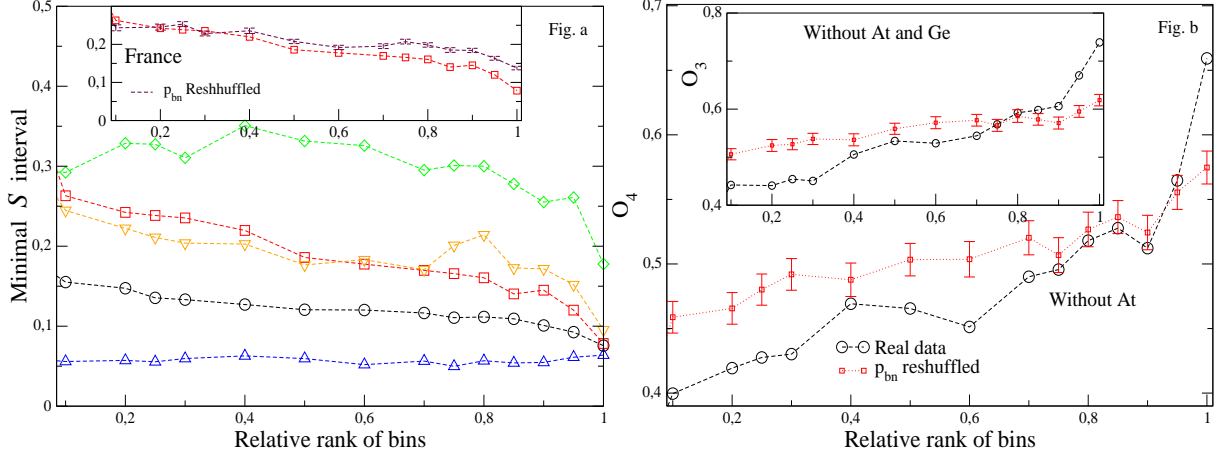


Figure 9. Quantitative evidence of the sharp and common peak of S for the most populated municipalities. Considered elections, the way that bins are ranked, countries or groups of countries and legends are the same as in Fig. 7. Left (9-a): Minimal interval ($S_{sup} - S_{inf}$), which encapsulates 50% of events, with respect to the relative population size. Right (9-b): Overlap O_4 between 4 distributions of S of municipalities as a function of their relative municipality-size (see text for the definition of O_n). The inset shows in the same manner overlap O_3 between 3 distributions of S ((1): all without At, Fr, Ge and Pl; (2): Fr; (3): Pl). Some curves obtained from reshuffling p_{bn} of municipalities (inside one country or ensemble of countries), while p_a is not modified, are also plotted.

What the common occurrence is not

We claim that this common most frequent value, $S \approx 1$ for the most populated municipalities, is not a mere statistical artefact. More precisely, we claim that:

- (1) it is not a direct consequence of the law of large numbers, which, for data aggregated at the scale of large municipalities, would give a systematic result;
- (2) it is not a result of ‘pure chance’, that is a bias in the data due to random events, or an accidental bias in the collected data;
- (3) it does not only result from having p_a and p_c neither around 50% nor around a common value: there is a wide range of p_a values for which $S \approx 1$ is observed;
- (4) it does not result from having a small proportion of Blank and Null Votes.

Let us now justify these claims.

• About the two first points

In support of the two first points, we note that there are robust properties which cannot be explained by the pure chance or the large number hypotheses. In particular:

- (i) $S \approx 1$ is specific to modern elections. Indeed (apart from Swiss *Votations* discussed in Section Discussion) this common value $S \approx 1$ appears recently, and at different times for different countries – and different elections –: in the 70’s or 80’s in France, 80’s in Germany, 90’s in Canada, 2000’s in Czech Republic, etc (cf. the SI, Fig. S1). Moreover, there is no systematic way in which recent convergence to $S \approx 1$ appears in time. $S \approx 1$ may be reached as well from inferior values (e.g. Chamber of Deputies elections in Canada, Czech Republic, etc., in Fig. 6-b) than from superior values (e.g. European Parliament in France in Fig. 6-b). Lastly, in a given country, some kind of elections provide at large scale

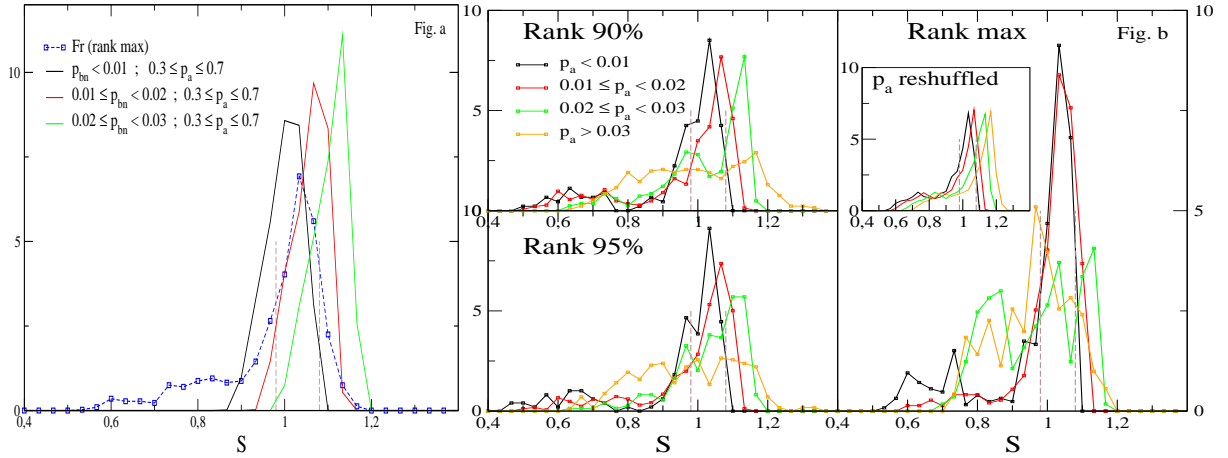


Figure 10. Relative importance of p_{bn} on the distribution of S . Left (10-a): Distribution of S from flat distributions of p_a (where $p_a \in [0.3; 0.7]$) and p_{bn} . The pdf of S can be peaked for a relatively broad distribution of p_a and a small range of p_{bn} , but the peak of the distribution of S , which depends on p_{bn} values, is not necessarily centered on $S \approx 1$. Histogram of S over the ≈ 100 French most populated municipalities (the same as in Fig. 7-d) is also given as a guiding view. Right (10-b): pdf of S , for different ranges of p_{bn} , over ≈ 100 French municipalities, which depend on their relative population-size (like in Fig. 7-b, c, d). Histograms of $S(p_a, p_{bn})$ from reshuffled p_a , while p_{bn} remain unchanged, are also given.

$S \approx 1$ since their coming (e.g. European Parliament elections), and for some other ones, $S \approx 1$ seems (actually) to be an attractor point in time (see e.g. Chamber of Deputies elections in Canada, Czech Republic, France, Switzerland, etc. in the SI, Fig. S1).

(ii) $S \approx 1$ is only observed for large populations (and there is no common-value for smaller municipality sizes) as it is shown in Fig. 9; and there is sometimes a plateau with a lower value of N which both depend on the election and on the country (e.g. ≈ 3000 in Canada and Czech Republic, 10000 in France for referendums, etc., in Fig. 4). Moreover, there is no systematic way in which convergence to $S \approx 1$ occurs as the population size increases. $S \approx 1$ may be reached as well from inferior values (e.g. Fr-1995-P2, Sp-2004-E and Sp-2009-E) than from superior values (e.g. Fr-2000-R, Ge-2004-E and Ge-2009-E in Fig. 4). Lastly, $S \approx 1$ may be reached from a discontinuous transition when voting rule (which depends on the population size of municipalities) changes. This occurs for the two French local elections for the Mayor (see Fig. 3), which are the only one elections of our database where there is this electoral rule change.

(iii) The shape of distributions of S for large municipality sizes does not result from a statistical bias due to large numbers: creating artificial high populated municipalities, by means of aggregating large amount of citizen choices who live in small and different municipalities, does clearly not yield a distribution peaked near $S \approx 1$ (see the SI, Section C and Fig. S6 for more details).

Finite-size-effects, that is the effect of aggregating data at different scales, are considered more thoroughly in the SI, Section C, comparing ballot box scale with municipality scale. This section also discusses more the issue of statistical effects that could be due to large numbers.

• About the two last points, concerning the ranges of p_a and p_{bn} values

We show that, even if the distributions of S could be peaked for a relatively broad distribution of p_a and small values of p_{bn} , this can not alone explain why the distributions of S for the most populated towns are so much narrowed and, in addition, have their peak at a common value of S .

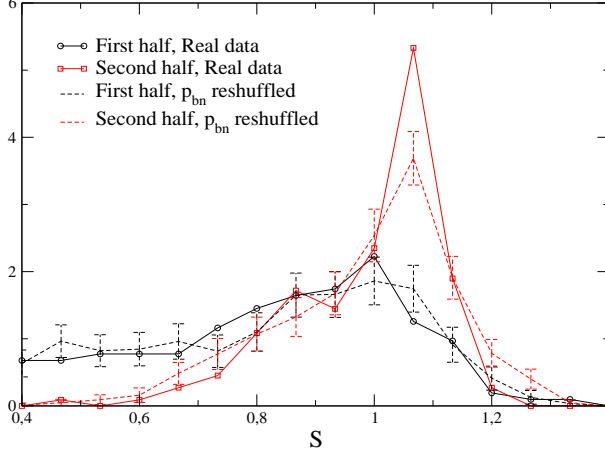


Figure 11. Relative importance of reshuffling p_{bn} on S . Analyzed elections and the manner that elections are divided into two groups are the same than in Fig. 5. Nevertheless, here, each election is aggregated at national scale, i.e. S is directly evaluated from the set $\{p_a, p_c, p_{bn}\}$ at the national scale. In these figures, surrogate $S(p_a, p_{bn})$ data, consist in reshuffling p_{bn} from one election to another one in the same group, while p_a is not modified. Surrogate curves result from the average of 1000 realizations, and standard-deviations are plotted as error bars.

Ranges of variation of p_a and p_{bn} .

On one side, while p_{bn} does not radically change in time at large scales, p_a has increased during last decades in most countries (see e.g. insets of Fig. 5 and the SI, Fig. S2). On the other side, p_{bn} is known to decrease when the population-size of municipalities N increases, as it was discussed in the Section Introduction. Let us thus first consider the possibility that the common occurrence $S \approx 1$ could be a consequence of these two facts: p_a is not too small⁸ (or p_c too small) and, *independently*, p_{bn} is small.

We give three arguments against this assertion. (i) First, we plot on Fig. 10-a histograms of S resulting from a flat and broad distribution of p_a , and a flat distributions of p_{bn} (with small values). Each histogram corresponds to a different choice of the range of (small) p_{bn} values.⁹ The result is indeed a set of peaked histograms. However, these distributions of S are neither necessarily centered on $S \approx 1$ nor centered at a common peak.

(ii) Second, we emphasize the specificity of most populated municipalities. Fig. 10-b plots for French data (where the tested phenomenon is clearer) distributions of S selecting elections for which p_{bn} belongs to specific ranges of values. Moreover these distributions are also plotted according to the population-size of municipalities. It is only for the most populated municipalities that the distributions of S for different ranges of p_{bn} are roughly peaked at the same value $S \approx 1$ (with a very good agreement for $p_{bn} \in [0, 0.01[$ and $p_{bn} \in [0.01, 0.02[$). Moreover, for a lower population-size, e.g. with a relative rank of 90%, it is interesting to note that distributions of S for different ranges of p_{bn} (apart from the $p_{bn} \geq 0.03$) share the same features as in Fig. 10-a, i.e. distributions are peaked in different values. (To have a more detailed view, see the SI, Fig. S5, which shows scatter-plots (p_a, p_{bn}) for the municipalities taken into account in Fig. 10-b.).

(iii) Third, there is actually a wide disparities in the ranges of p_a and p_{bn} between different countries or group of countries. One can see in Fig. 8 how, (1), France and, (2), all countries without At, Fr, Ge and Pl, can reach the common S peak, despite largely different ranges of p_a , p_c and p_{bn} . In other words, the ranges of ratios p_a , p_c and p_{bn} are not sufficiently similar between countries or ensemble of countries to explain why the distributions of S for the most populated municipalities share a sharp peak at a common value of S .

Implied correlations between p_a and p_{bn} .

Hence, it seems difficult to explain the common value $S \approx 1$ for the most populated towns as a consequence of having *independently* p_{bn} small and p_a in a given particular range. The observation of a common peak

⁸For example, if $p_a \lesssim 0.227$, then it is no more possible to get $S = 1$.

⁹To better understand this point, let $S_2(p_a) = -p_a \log(p_a) - (1 - p_a) \log(1 - p_a)$ which has a maximal value, $S_2 = 1$, for $p_a = 0.5$. Moreover, when $p_{bn} = 0$, S_2 is equal to the involvement entropy, S (defined in Eq. (2)), i.e. $S_2(p_a) = S(p_a, p_{bn} = 0)$. Hence, relatively small variations of p_a around 0.5 and very small values of p_{bn} lead to $S \approx 1$.

around $S \approx 1$ thus implies the existence of specific correlations between p_a and p_{bn} .

To test this conclusion, we consider surrogate data obtained by reshuffling the ratios p_{bn} from one municipality (or country) to another one, while p_a is kept unchanged (and then p_c is deduced from $p_c = 1 - p_a - p_{bn}$). Note that the marginal distributions of p_a and p_{bn} remain unchanged by this reshuffling procedure, whereas their correlations are destroyed. We use this method twice: first, (i) contrasting recent and old elections, and second, (ii), considering the dependency in municipality size.

(i) Figure 11 shows, at national scale and for two periods of time, how the distributions of S change under this reshuffling. p_{bn} are reshuffled within the same group of elections. For the first period in time, the real distribution of S , which is not peaked near $S \approx 1$, and the surrogate one are not very different between themselves. By contrast, the distributions are notably different for the second period. Moreover, the main difference concerns the peak near $S \approx 1$. The peak of the surrogate data distribution is less sharp than the one of the real data. This is particularly interesting since p_{bn} is roughly distributed in the same manner between the two relative periods in time (see insets of Fig. 5 or scatter-plots (p_a, p_{bn}) in the SI, Fig. S4). The widening of the surrogate distribution of involvement entropy near the peak $S \approx 1$ can be seen as a sign that there are correlations between p_a and p_{bn} which enforces the occurrence of $S \approx 1$.

(ii) From a qualitative point of view, the reshuffled data have a peak of S values which is less narrow than for the real ones, a discrepancy which increases with municipality size, as can be seen for the French data on the inset of Fig. 9-a, and on the scatter-plots (p_a, p_{bn}) on Fig. S5 of the SI. In addition, the distributions of S obtained for the reshuffled data are not as well peaked at a common value as it is the case for the real data ones. Quantitatively, for the French data, the Kolmogorov-Smirnov distance between the distributions of real and reshuffled data is significantly larger for the most populated municipalities, with a distance that allows one to reject the hypothesis that the two distributions are similar (indeed the Kolmogorov-Smirnov distance is then 3.0 ± 0.2 , while 1.6 corresponds to $\approx 1\%$ probability that the two distributions coincide). Moreover, Fig. 9-b shows that overlaps between different distributions of S resulting from reshuffled p_{bn} is smaller than for real data, and this only when municipality-sizes are high, or even only for the most populated municipality sample: the reshuffling suppresses the high increase of overlaps which is observed on real data for the sample of the most populated municipalities.

We can thus conclude that there is a specific property for the most populated municipalities, which is not encapsulated by considering p_a and p_{bn} as independent variables.

Discussion

We suggest that the common value $S \approx 1$ of the entropy, which appears recently in high populated municipalities, reveals an emerging collective behavioral norm characteristic of citizen involvement in modern democracies, and we propose to call it a ‘weak law’ on recent electoral behavior among urban voters. Signs of existence of this possible norm can not only be seen notably by the greatest density value of the involvement entropy S around ≈ 1 , whatever countries, type of elections, etc., but also by its deviances. There are two kinds of deviances: for the first one, S is small (which generally occurs when p_a or p_c is very small), for the second one, S is high (which generally results from great ratio of blank or null votes, p_{bn}). We will see that these deviances are associated with a particular phenomena of civic involvement, or are simply reduced to the norm (i.e $S \approx 1$) when the meaning of blank votes changes.

When significantly smaller values are observed (e.g. $S \lesssim 0.85$) for cities, something appear inside towns (in average): the heterogeneity of involvement entropy over all polling stations of a given town decreases when S of the whole city decreases. In other words, considering the electorate civic involvement in a given town, the less is S for the whole town, the more the town appears homogeneous (i.e. involvement entropies, at polling station scale, over all polling stations of the town are more homogeneous between themselves). Section E of the SI shows this point (free of statistical bias), particularly clear when the ratio p_c is high (compared to cases where p_a are high). This civic involvement phenomenon for towns

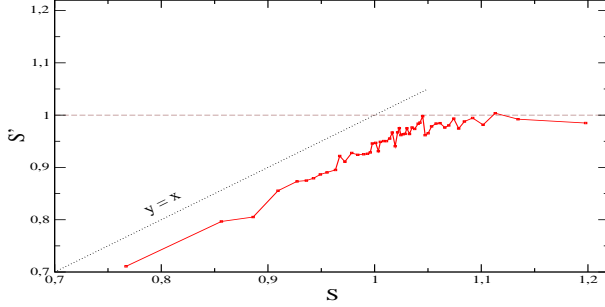


Figure 12. A modified involvement entropy, S' , where Blank Votes are grouped with Valid Votes, with respect to the involvement entropy S , for ≈ 500 Swiss referendums. (See in the SI, Section F for a deeper discussion.) Each point corresponds to the average of about 10 referendums. Note the plateau $S' \approx 1$ for $S \gtrsim 1.05$.

with small S can be seen as a signature of something ‘new’ which appears when deviance of the norm occurs.

On the other hand, elections where significantly $S > 1$, typically corresponds to cases where there has been an appeal (from political parties, citizens blogs, etc.) to vote blank or null, which adds civic-involvement ‘tensions’ to the election. It is remarkable that countries which make the distinction between blank votes to null votes, provide, by considering blank votes like the valid votes in favor of one of the list of choices, a modified involvement entropy $S' \approx 1$ whenever the involvement entropy is $S > 1$. (When blank votes are grouped with votes according to the list of choices, the modified involvement entropy S' is equal to $S(p_a, p_c + p_b, p_n)$ in Eq. (2), and not $S(p_a, p_c, p_b + p_n)$ as for the usual involvement entropy, where p_b , p_n and $p_{bn} = p_b + p_n$ mean respectively ratios of blank votes, null votes and blank or null votes.) See the striking plateau in Fig. 12 for Swiss referendums, which shows a modified involvement entropy $S' \approx 1$ when $S > 1$. Moreover, Section F of the SI clearly shows this point, e.g. for European Parliament elections in Italy, and for Referendums in Spain. Hence the fact that $S > 1$ boils down to a modified involvement entropy ≈ 1 , by categorizing blank votes as Valid Votes, can be seen as the recovering of the ‘weak law’ by the decrease of civic involvement ‘tensions’. The fact that a deviance of the norm is naturally reduced to the norm (the involvement entropy is around 1) as soon as blank votes are grouped with ‘valid votes’ can be seen not as an haphazardly occurrence but rather as a signature of the norm in a larger sense.

Now let us discuss a more about the term ‘weak law’. In one hand, the common value $S \approx 1$ (for the most populated municipalities in a recent times) appears as a kind of law of a phenomenon not yet measured up to now. This phenomenon concerns the involvement of the electorate, in a civic point of view. A kind of law, because it occurs very frequently, without being based on a ‘pure chance’ phenomenon, despite wide disparities across elections, with strong regularities, and, as we have seen, it implies the existence of particular correlations between p_a and p_{bn} . In other hand, this is clearly not a ‘hard law’ since strong deviations are still existing. One cannot exclude that a larger law exists, encapsulating more regularities for the most populated municipalities (e.g. by taking into account the political context, number of valid votes for different choices, etc.), and which might explain why $S \approx 1$ appears in recent time and do not usually concern small municipalities. It also may be the case that another function than the civic involvement entropy encapsulates more regularities about the set of ratios $\{p_a, p_c, p_{bn}\}$. In any case, we believe that this weak law of recent urban civic involvement shows up as a consequence of some robust electoral behavior. As one more illustration, Swiss referendums show (at the *canton* scale) small fluctuations of S near this same value, $S \approx 1$, from 1880s to nowadays (see Fig. 13).

To conclude, the main finding of this work, based on the analysis of a wide number of elections from 11 different countries, is that a common stylized fact emerges: in recent elections, the distribution of the involvement entropy is found to be sharply peaked near $S \approx 1$, in high populated municipalities (and thus also at national levels). This universal property is remarkable given the wide disparities across countries

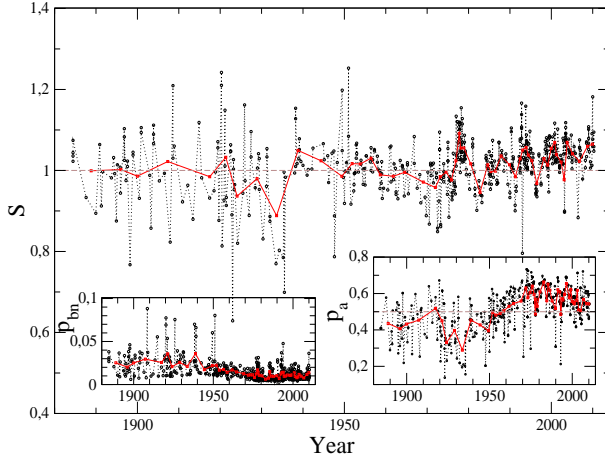


Figure 13. Time evolution of the involvement entropy, S , of 531 Swiss referendums, at large scale. Each point corresponds to the average (weighted by the number of registered voters) over all Swiss *cantons* (25 or 26 in quantity). In red (as a guide view): average values over ≈ 10 referendums. The inset show same things, but for ratio of abstentionists, p_a , and the ratio of blank and null votes, p_{bn} .

(and even within countries for different elections) in political mores, voting systems, in the way that lists of registered voters are established (on a voluntary basis or automatically, etc.), and so on.

Moreover, $S \approx 1$ appears to be very stable in time whenever it occurs for one kind of election, as for example European Parliament elections in Western Europe, and particularly remarkably for the Swiss referendums since 1884. We propose to designate this strong regularity, neither a ‘hard law’ nor a mere statistical artefact, as a ‘weak law’ of electoral involvement characteristic of modern democracies in urban cities. We suggest that the existence of this weak law is the signature of an emerging collective behavioral norm. More studies and analysis would be necessary in order to better understand its conditions of realizations and its meaning (at the individual scale and/or at macro scale). Moreover, it should be very interesting for forthcoming studies, notably to know if this ‘weak law’ also occurs in emergent countries, in new democratic countries, in great cities (whatever they are), etc.

The present study calls for a different point of view than those commonly used in Political Sciences. We do not work within the classical paradigm explaining the electoral behavior with sociological or ethnic even institutional or rational choice variables. Our propose is to change perspective of observation, using very large sets of data, looking for regularities – stylized facts –, without restricting the analysis to a particular category which could be based on chronology, space, institutional or national specificities. At a ‘macro’ level, using aggregated data, and not at the individual scale, this new view point focuses on (1) the involvement or the mobilization of the electorate, and (2) a measure of heterogeneity or, otherwise stated, of order and disorder. The question asked here to electoral data is not why a more or less rational citizen participates or not to an election, but *how is the degree of disorder of civic involvement of the electorate*.

Materials and Methods

The SI Section A gives more information about the set of (public) electoral data studied in this paper. Most of them can be directly downloaded from official websites (see the SI, References). Part of the database used in this paper can also be directly downloaded from [26].

Average values and standard-deviations do not take into account extreme values in order to remove some electoral errors, etc. Electoral values greater than 5σ (or 3σ for the polling stations of a town, as in the SI Section E) are not taken into account.¹⁰

¹⁰For instance let 100 municipalities of size $\approx N$ (as in Fig. 4), each one has a civic involvement entropy S_i ($i = 1, 2, \dots, 100$). First, $\langle S \rangle$ and σ are the average value and the standard-deviation of S over these 100 municipalities. Next, the final average value \bar{S} and the final standard-deviation over this sample of 100 municipalities are uniquely evaluated for municipalities, i ,

Curves resulting from reshuffling procedure give the average values over 1000 realizations, and standard-deviations are plotted as error bars.

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Between order and disorder: a ‘weak law’ on recent electoral behavior among urban voters?

Christian Borghesi, Jean Chiche and Jean-Pierre Nadal

Supporting Information

A. Data

• Elections studied at municipality scale

Table S1 gives more details about the 76 elections studied in this paper at the municipality scale. There are: 13 elections from Austria [1] (≈ 2400 municipalities)¹¹; 5 from Canada [2] (≈ 7700 municipalities); 1 from Czech Republic [3] (≈ 6400 municipalities); 20 from Metropolitan France [4] (≈ 36000 municipalities); 7 from Germany [5] (≈ 12000 municipalities)¹²; 4 from Italy [6] (≈ 8100 municipalities); 4 from Mexico [7] (≈ 2400 municipalities)¹³; 11 from Poland [8] (≈ 2500 municipalities)¹⁴; 4 from Romania [9] (≈ 3200 municipalities)¹⁵ ¹⁶, 4 from Spain [10] (≈ 8100 municipalities) and 3 from Switzerland [11] (≈ 2700 municipalities)¹⁷.

Table S1 also gives basic statistics over the ≈ 100 (≈ 200 for France) most populated municipalities.

• Time evolution at the national or provincial scale

The study of time evolution of S is done for the same countries as in Tab. S1 and for all national elections for which we have enough data. For Austria [12], the study considers data since 1945, even if compulsory voting was abolished in the whole country in 1992 for National Council elections (D), and after 2004 for Presidential elections (P) (but in 1982 some provinces had yet done it); for Canada [13], since 1945; for Czech Republic [14], since 1990¹⁸; for France [15], since 1945¹⁹; for Germany [16], since 1949; for Italy [17], since 1945 even if there were compulsory voting until 1993²⁰; for Mexico [18], since 1991; for

¹¹Corrections due to *wahlkarten* or postal votes are taking account from the national level, i.e. in this paper, each municipality receive from voting cards a number of votes and valid votes proportional to its number of population, and at the same ratio for every municipality.

¹²Chamber of Deputies (D) elections refer to the *German Bundestag* elections. *Land* Parliament elections at time less or equal to 2004 (or 2010) in each *Land* are written here as ‘2004 Ld’ (or ‘2010 Ld’). Postal votes (*brichwahlen*) are usually taken account at *Landkreis* scale (they are distributed in municipalities, according to their populations), when it is possible to do it. Nevertheless, these corrections provide a very small difference in Fig. 4, especially for high population-size bins.

¹³The 2006 *Senador* election (not studied here) gives a very near statistics of S than the (P) and (D) elections that also occur at the same time.

¹⁴The Chamber of Deputies (D) election is the *Sejm* Chamber election.

¹⁵The referendum studied here is about *the reduction of the number of parliamentarians to a number of 300 persons*, and not about the *adoption of a unicameral Parliament* held on the same time. The latter one is not known at the polling station level.

¹⁶Some Romanian electors, not registered in the *lista electorala permanenta*, are able to vote. For this country, we pursue to write N the Number of Register Voters, N_v the registered electors who take part to the election, and N_{bn} the number of Null and Blank Votes that the Registered Voters could make (even if the latter data is not known.) Romanian electoral data gather for each municipality, N , N_v , $N_v(tot)$ (the total number of votes), and $N_{bn}(tot)$ the total number of Null and Blank votes. Assuming that registered electors and not registered electors vote Null and Blank in the same way (i.e. $\frac{N_{bn}(tot)}{N_v(tot)} = \frac{N_{bn}}{N_v}$), we deduce N_{bn} .

¹⁷The referendums or *vatations* ‘R(a)’ and ‘R(b)’ respectively occurred the 11 of March and the 17 of July. The Legislative (D) election refers to the *Conseil National* election.

¹⁸The 1990 and 1992 Deputies (D) elections only refer to the Parliamentary Chamber of People election. The Parliamentary Chamber of Nations and the Parliamentary National Council elections, that occurred at the same day as the previous ones, also gave approximately the same S value.

¹⁹All French electoral data are from metropolitan France. Some referendums are not known at the *département* scale. In these cases, S is evaluated at the national scale.

²⁰We consider the only first question asked to electors in referendums.

Id	\overline{S}	$\overline{p_a}$	$\overline{p_c}$	$\overline{p_{bn}(\overline{p_b})}$	Id	\overline{S}	$\overline{p_a}$	$\overline{p_c}$	$\overline{p_{bn}(\overline{p_b})}$
Fr 1992 R	1.02 \pm 0.04	0.32	0.66	0.018	Fr 1993 D	1.09 \pm 0.04	0.34	0.63	0.028
Fr 1994 E	1.12 \pm 0.03	0.48	0.50	0.020	Fr 1995 P1	0.91 \pm 0.04	0.24	0.74	0.018
Fr 1995 P2	1.01 \pm 0.07	0.23	0.73	0.044	Fr 1997 D	1.08 \pm 0.03	0.36	0.62	0.024
Fr 1998 rg	1.11 \pm 0.03	0.46	0.52	0.019	Fr 1999 E	1.11 \pm 0.03	0.54	0.44	0.020
Fr 2000 R	1.02 \pm 0.07	0.71	0.25	0.036	Fr 2002 P1	1.01 \pm 0.04	0.31	0.67	0.019
Fr 2002 P2	0.95 \pm 0.07	0.21	0.75	0.035	Fr 2002 D	1.02 \pm 0.04	0.37	0.62	0.010
Fr 2004 rg	1.10 \pm 0.04	0.41	0.57	0.021	Fr 2004 E	1.04 \pm 0.03	0.57	0.42	0.010
Fr 2005 R	1.00 \pm 0.05	0.32	0.66	0.014	Fr 2007 P1	0.72 \pm 0.08	0.17	0.82	0.010
Fr 2007 P2	0.84 \pm 0.06	0.17	0.80	0.032	Fr 2007 D	1.04 \pm 0.03	0.42	0.57	0.009
Fr 2009 E	1.03 \pm 0.05	0.60	0.39	0.012	Fr 2010 rg	1.06 \pm 0.03	0.57	0.42	0.012
At 1994 D	0.81 \pm 0.11	0.20	0.78	0.016	At 1995 D	0.73 \pm 0.10	0.15	0.83	0.018
At 1996 E	1.04 \pm 0.04	0.33	0.65	0.021	At 1998 P	1.00 \pm 0.10	0.27	0.70	0.032
At 1999 E	1.06 \pm 0.05	0.52	0.46	0.013	At 1999 D	0.82 \pm 0.09	0.22	0.77	0.011
At 2002 D	0.73 \pm 0.10	0.17	0.81	0.011	At 2004 P	1.04 \pm 0.09	0.31	0.66	0.028
At 2004 E	1.03 \pm 0.05	0.59	0.40	0.010	At 2006 D	0.87 \pm 0.09	0.24	0.74	0.012
At 2008 D	0.88 \pm 0.08	0.24	0.75	0.014	At 2009 E	1.04 \pm 0.04	0.55	0.44	0.009
At 2010 P	1.16 \pm 0.06	0.48	0.49	0.034					
Pl 2000 P1	0.98 \pm 0.03	0.36	0.63	0.006	Pl 2001 D	1.09 \pm 0.02	0.52	0.46	0.015
Pl 2003 R	0.98 \pm 0.02	0.37	0.62	0.004	Pl 2004 E	0.79 \pm 0.07	0.78	0.22	0.005
Pl 2005 D	1.06 \pm 0.03	0.58	0.41	0.013	Pl 2005 P1	1.02 \pm 0.01	0.49	0.51	0.003
Pl 2005 P2	1.03 \pm 0.01	0.47	0.53	0.006	Pl 2007 D	1.05 \pm 0.03	0.42	0.57	0.010
Pl 2009 E	0.87 \pm 0.06	0.73	0.27	0.004	Pl 2010 P1	1.01 \pm 0.02	0.43	0.57	0.004
Pl 2010 P2	1.03 \pm 0.02	0.43	0.56	0.007					
Ge 2002 D	0.83 \pm 0.07	0.22	0.77	0.009	Ge 2004 Ld	1.02 \pm 0.04	0.41	0.58	0.007
Ge 2004 E	1.02 \pm 0.05	0.59	0.40	0.009	Ge 2005 D	0.87 \pm 0.06	0.24	0.75	0.011
Ge 2009 E	1.00 \pm 0.05	0.60	0.40	0.006	Ge 2009 D	0.95 \pm 0.05	0.30	0.69	0.009
Ge 2010 Ld	1.04 \pm 0.03	0.43	0.56	0.009					
Ca 1997 D	1.00 \pm 0.04	0.37	0.62	0.009	Ca 2000 D	1.03 \pm 0.03	0.44	0.56	0.006
Ca 2004 D	1.02 \pm 0.02	0.46	0.54	0.004	Ca 2006 D	1.01 \pm 0.02	0.44	0.56	0.003
Ca 2008 D	1.02 \pm 0.02	0.49	0.51	0.003					
It 2004 E	1.11 \pm 0.12	0.29	0.66	0.053(0.023)	It 2006 D	0.78 \pm 0.13	0.17	0.81	0.020(0.007)
It 2008 D	0.89 \pm 0.12	0.20	0.77	0.027(0.008)	It 2009 E	1.08 \pm 0.10	0.36	0.61	0.034(0.013)
Mx 2003 D	1.04 \pm 0.05	0.59	0.40	0.013	Mx 2006 D	1.04 \pm 0.04	0.40	0.58	0.012
Mx 2006 P	1.03 \pm 0.04	0.40	0.59	0.010	Mx 2009 D	1.11 \pm 0.06	0.56	0.41	0.027
Ro 2009 E	0.73 \pm 0.09	0.81	0.18	0.008	Ro 2009 R	1.09 \pm 0.02	0.55	0.44	0.017
Ro 2009 P1	1.05 \pm 0.02	0.52	0.48	0.008	Ro 2009 P2	1.04 \pm 0.02	0.50	0.50	0.006
Sp 2004 D	0.92 \pm 0.07	0.24	0.74	0.020(0.014)	Sp 2004 E	1.01 \pm 0.06	0.57	0.42	0.006(0.003)
Sp 2008 D	0.91 \pm 0.08	0.26	0.73	0.013(0.009)	Sp 2009 E	1.03 \pm 0.04	0.56	0.43	0.009(0.006)
CH 2007 R(a)	1.04 \pm 0.04	0.53	0.46	0.008(0.004)	CH 2007 R(b)	0.99 \pm 0.06	0.62	0.37	0.007(0.004)
CH 2007 D	1.04 \pm 0.05	0.53	0.47	0.009(0.002)					
Cz 2003 R	1.07 \pm 0.01	0.47	0.52	0.012					

Table S1. Elections studied in this paper at the municipality scale. An election is identified (Id) by its country, its year date and its nature. D: Chamber of Deputies election; E: European parliament election; P: presidential election (according to the constitution of the country, in only one round); P1 and P2: first and second round of a Presidential election; R: Referendum; Ld: German *Länder* elections; rg: French *Régionales* elections. For each country elections are given in a chronological order (but the 2006 Mexican Presidential (P) and Deputies (D) elections occurred the same day, and also for the 2009 Romanian Presidential (P1) and Referendum (R) elections). Even if an election needs two rounds, only the first one is considered (e.g. the French Deputies (D) and *Régionales* (rg) elections) unless the contrary is indicated (e.g. P1 and P2). Mean values of S , p_a , p_c , p_{bn} (and (p_b) if Blank Vote are distinguished between Null Vote), and also standard deviation only for S , are given over the bin of the ≈ 100 (or ≈ 200 for France only) most populated municipalities. In bold text, $\overline{S} \in [0.98; 1.08]$.

Country	Kind of elections	Scale of aggregate data
At	D, E, P, R	National
Ca	D	Province (5-13)
CH	D, R	<i>Canton</i> (25-26)
Cz	D, E, R, rg, S1, S2	National
Fr	Cant, D, E, P1, P2, R, rg	<i>département</i> (90-96)
Ge	D, E	<i>Land</i> (9-16)
It	D, E, R, S	National
Mx	D, P	National
Pl	D, E, P1, P2	National
Ro	D, E, P1, P2, R	National
Sp	D, E, R	<i>Comunidad autónoma</i> (17-19)

Table S2. Elections studied in this paper for their global S as a function of time. Notation is the same as in Tab. S1. For Czech Republic, “rg” means Election into regional councils, “S1” and “S2” are respectively the first and second round of the Senate elections; for France, “Cant” refers to the *Cantonaes* elections and some referendums are only known at the national scale; for Italy, “S” means Senate elections, and occur at the same time as Deputies elections (D) but with older registered voters. In parenthesis, the total number of different provinces (or *Cantons*, etc.), which can change in time, in the whole country.

Poland [19], since 1990 ²¹; for Romania [20], since 1990; for Spain [21], since 1976; for Switzerland [22], since 1884 for referendums (R) and since 1919 for legislative elections (D). If an election needs two rounds, the first one is considered, unless the contrary is indicated. The Mexican, Polish and Romanian Senate elections are not shown here because they occur at the same time as Chamber of Deputies elections and have very similar S results.

Table S2 summarizes the nature of elections studied in this paper, and also the scale of aggregate data per country. Note that the last election analyzed in this paper is the Referendum which held in Italy on June 2011. ²²

• Elections studied at polling station scale

Polling stations analysis is restricted to polling stations which belong to one of the 100 most populated municipalities (for the considered election). 31 elections at the polling station scale are studied in this paper: 5 for Canada (each Canadian election of Tab. S1), with around 25000 polling stations; 13 for France (French elections of Tab. S1 since 1999), with around 7000 polling stations; 4 for Mexico (each Mexican election of Tab. S1), with around 55000 polling stations or ballot box; 5 for Poland (Polish election of Tab. S1 from 2003 up to 2005), with around 8000 polling stations; and 4 for Romania (each Romanian election of Tab. S1), with around 6000 polling stations. See Tab. S3 for some basic statistics over polling stations of the 100 most populated municipalities.

²¹We have not data from the 1989 Chamber of Deputies (Sejm) election nor the two referendums in 1996.

²²Official results (which took into account registered voters) of the Canadian Chamber of Deputies election, held on May 2011, were not published at the time we first submitted this paper. In Fig. S1, the involvement entropy over all provinces would be $\approx 1.00 \pm 0.02$ and respectively 0.99 and 1.02 for Ontario and Quebec.

Id	S	τ_3	Id	S	τ_3
Fr 1999 E	1.09 ± 0.05	-3.7 ± 0.6	Fr 2000 R	1.00 ± 0.11	-4.2 ± 0.6
Fr 2002 P1	1.00 ± 0.06	-2.1 ± 0.6	Fr 2002 P2	0.93 ± 0.10	-0.7 ± 0.6
Fr 2002 D	1.01 ± 0.06	-3.2 ± 0.7	Fr 2004 rg	1.09 ± 0.05	-2.8 ± 0.6
Fr 2004 E	1.03 ± 0.05	-4.6 ± 0.7	Fr 2005 R	0.99 ± 0.07	-2.5 ± 0.7
Fr 2007 P1	0.71 ± 0.11	-1.3 ± 0.7	Fr 2007 P2	0.83 ± 0.09	-0.1 ± 0.7
Fr 2007 D	1.03 ± 0.04	-3.8 ± 0.7	Fr 2009 E	1.02 ± 0.07	-4.6 ± 0.7
Fr 2010 rg	1.04 ± 0.05	-4.4 ± 0.7			
Ca 1997 D	0.98 ± 0.08	-3.3 ± 1.3	Ca 2000 D	1.00 ± 0.06	-4.1 ± 1.1
Ca 2004 D	1.00 ± 0.05	-4.4 ± 0.9	Ca 2006 D	0.99 ± 0.05	-4.4 ± 0.8
Ca 2008 D	1.00 ± 0.05	-4.6 ± 0.9			
Pl 2003 R	0.95 ± 0.10	-4.0 ± 0.9	Pl 2004 E	0.83 ± 0.13	-6.3 ± 0.8
Pl 2005 D	1.05 ± 0.08	-4.1 ± 0.9	Pl 2005 P1	1.00 ± 0.07	-4.9 ± 0.9
Pl 2005 P2	1.01 ± 0.05	-4.1 ± 0.9			
Mx 2003 D	1.03 ± 0.07	-4.3 ± 0.9	Mx 2006 D	1.02 ± 0.07	-3.2 ± 0.8
Mx 2006 P	1.01 ± 0.07	-3.4 ± 0.8	Mx 2009 D	1.11 ± 0.10	-3.5 ± 0.9
Ro 2009 E	0.70 ± 0.13	-6.6 ± 0.9	Ro 2009 R	1.08 ± 0.05	-3.8 ± 0.7
Ro 2009 P1	1.04 ± 0.03	-4.5 ± 0.7	Ro 2009 P2	1.04 ± 0.03	-4.4 ± 0.7

Table S3. Elections studied at the polling station level. An election is identified (Id) by its country, its year date and its nature. Mean value and standard deviation of S and of τ_3 (see the SI Section D) over ballot boxes in the 100 most populated municipalities.

B. More details on data analysis

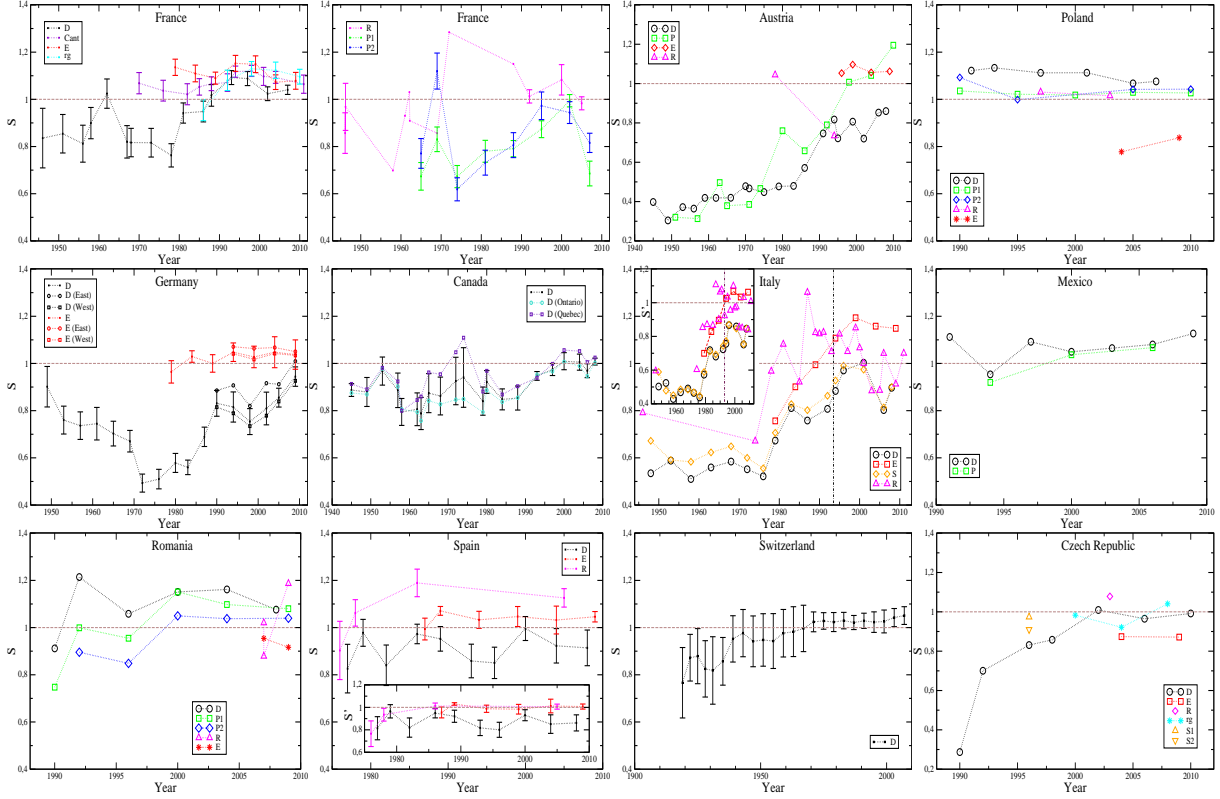


Figure S1. Time evolution of the mean involvement entropy at large scale (national, provincial, etc.). See Section A and Tab. S2, for more details and also for the end of compulsory voting in Italy (cf. vertical dashed line) and in Austria. Whenever the scale of aggregate data is lower than the national one, standard-deviations (weighted by the number of registered voters) are also shown as error bars. Italian and Spanish graph insets show a variant of S where Blank Votes are categorized as Valid Votes (see Section F for more discussion). See text for more explanation about some French curves.

Fig. S1 gathers all the available data (see in the SI, Section A for more details) at a large aggregate scale (country, province, *département*, etc.). When the scale of aggregate data is lower than the national one, each point corresponds to a weighted (by population-size) mean value of involvement entropies at lower scale (province, *département*, etc.), and standard deviation is also given as error bar. The cases where Blank Votes are distinguished from Null Votes (i.e. in Italy, Spain and Switzerland), call for a specific discussion (see the SI, Section F).

Let us comment Fig. S1 on the case of the Chamber of Deputies elections in France, at the large scale called *département* (96 in quantity for metropolitan France, actually). One sees an involvement entropy frequently equal to ≈ 0.8 until 1981, which then increases and gets greater than 1 until 2000, and decreases a little and stabilizes to $S \approx 1$ after 2000. So, the civic involvement of the electorate (at the *département* scale) is relatively ordered until 1981 and get more and more disordered until 2000. After 2000, S seems to stabilize to a common value $S \approx 1$ which is also reached for the European Parliament elections and for local elections at different scales, such as the *Régionales* (\sim states) and the *Cantonales* (\sim counties) elections.

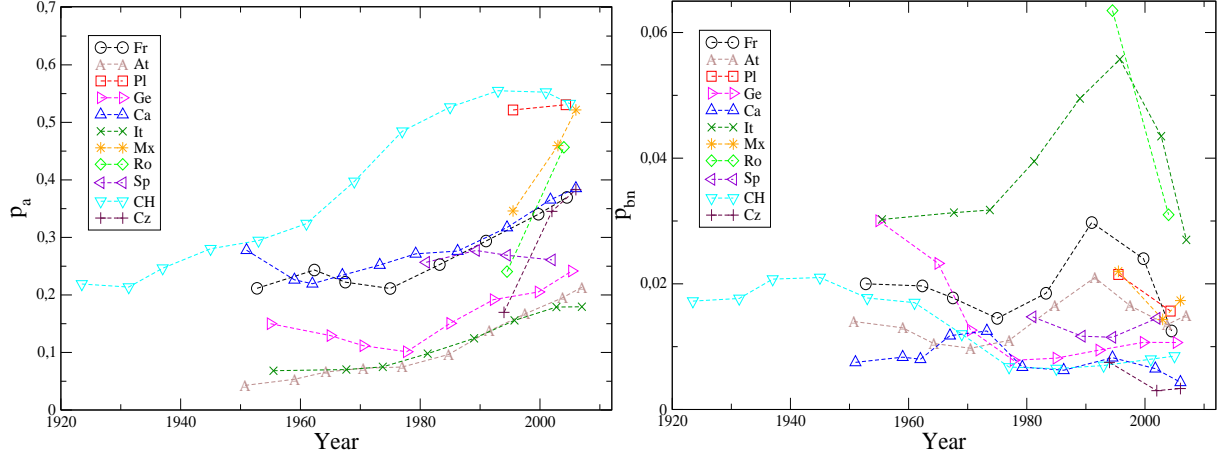


Figure S2. Moving average, as a function of time, per country of p_a and p_{bn} at national scale for Chamber of Deputies elections. The average is made over 4 elections. Left: about ratio of registered voters who do not take part to the election (p_a); Right: about Blank and Null ratio (p_{bn}).

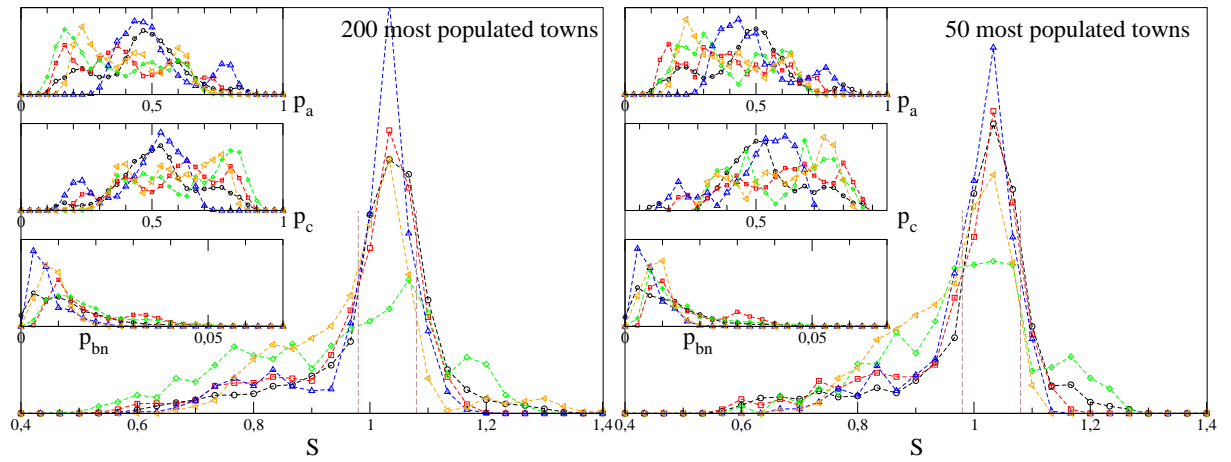


Figure S3. Histograms of S for the ≈ 200 (left) and 50 (right) most populated municipalities, similarly to Fig. 7-d (with 100 most populated municipalities for the latter one).

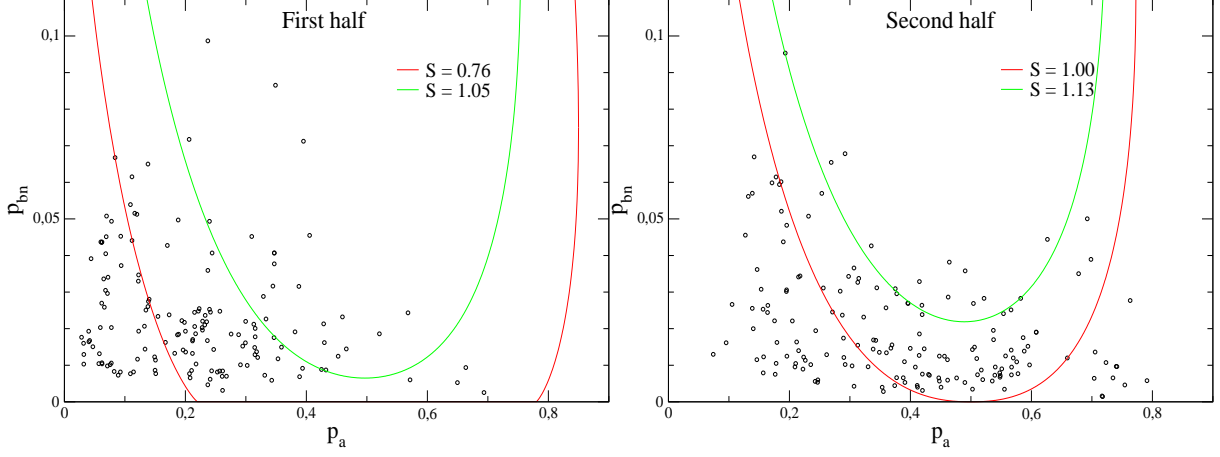


Figure S4. Evolution in time of scatter plots of (p_a, p_{bn}) at national level of 321 elections. Elections are divided into the two groups in the same manner as in Fig. 5. Curves give the sets of points (p_a, p_{bn}) such that $S(p_a, p_{bn})$ is equal to one of the two endpoints of the minimal interval of S which contains 50% of events. Note if S is equal to the average value (weighted by the population size) at lower aggregate scale (as provinces, *départements*, etc.) like in Fig. 5, the peak of S near $S \approx 1$ would be more narrowed and more centered on $S = 1$

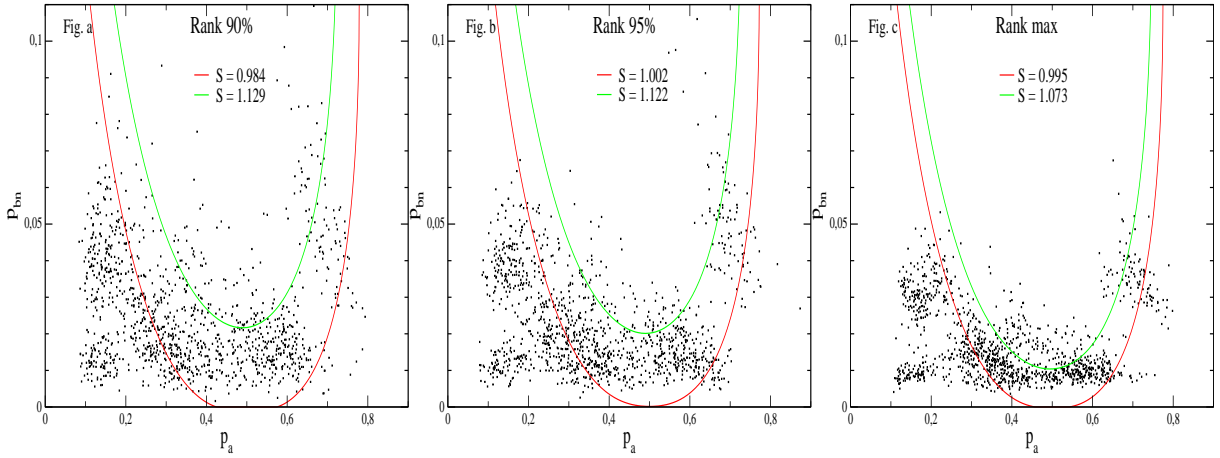


Figure S5. Scatter plots of (p_a, p_{bn}) of French municipalities according to their relative population size, over elections since 2000 (similarly as in Fig. 7-b, c, d). The sets of points (p_a, p_{bn}) such that $S(p_a, p_{bn})$ is equal to one of the two endpoints of the minimal interval of S which contains 50% of events (as in Fig. 8 for the most populated municipalities) are also plotted.

C. Finite size effects

We show in this section that finite size effects over municipality-size, N , on the entropy-involvement S , are relatively small for the most populated municipalities. Biases due to finite size effects have two possible origins: (1) level of aggregation of the data, over N about a hundred to a million, influences S measures, and (2) a statistical effect due to large numbers. Without a loss of generality, we examine these two biases for French electoral data – with 20 elections at the municipality scale and 13 at the polling station level, cf. the SI Section A. Lastly we show that the distribution of the involvement entropy which is sharply peaked near $S \approx 1$ for most populated towns is not due to considering a large number of N per town.

(1) Scale at which data are aggregated

French municipality sizes range from around 10 to around 100,000. In order to investigate how aggregate data scale modifies the measurement of the involvement entropy S , for each municipality we compare the results at the municipality scale with the one done at the polling station scale. Registered voters per polling station do not exceed around one thousand in France. We compare for a municipality its involvement entropy, S , measured at the municipality level, to the mean value, S_{PS} , of the involvement entropy over all the polling stations in the considered municipality. Convexity of the logarithmic function implies that the later is at most equal to the former. For each of the 200 most populated French municipalities, and for each of the 13 French elections known at the polling station scale (see the SI Section A), the gap between S and S_{PS} is less than about 2% (except for very few and typical recording errors of electoral data). Moreover, averaging S and S_{PS} over samples of ≈ 200 municipalities of similar sizes N provides a difference less than 1% for $N \gtrsim 1000$.

In short, for large population municipalities, the bias introduced by the scale at which data are aggregated is weak and does not affect the main conclusions of the paper.

(2) Statistical effects due to large numbers

Let us see if statistical fluctuations due to finite size effects considerably modify the expected values of involvement entropy. Indeed, For independent events, according to the central limit theorem (under conditions broadly applicable) fluctuations are on the order of $1/\sqrt{N}$. This is expected to be the case for the ratios p_a and p_{bn} , which should then lead to a bias in the entropy value. We want to estimate this bias and see if it is negligible (say less than 1%). To do so, we make a simulation with artificial data. For calibrating these data, we make use of the sample of the most populated municipalities. We measure the average values $\overline{p_a}$ and $\overline{p_{bn}}$ of p_a and p_{bn} over all municipalities in this sample of the largest municipality-size; and the corresponding standard deviations σ_a and σ_{bn} . The surrogate data consists in a same number of “municipalities”, each one characterized by the same population size as in the empirical data. For these surrogate-municipalities, we draw the numbers of Abstentionists and of Null-Blank votes from binomial distributions, parametrized by the empirical average values and standard deviations of p_a and p_{bn} , as follows.

Let a surrogate-municipality with N registered voters. Its numbers of Abstentionists, N_a , and Null-Blank votes, N_{bn} , are drawn from a binomial distribution such that:

$$\begin{aligned} N_a &= \mathcal{B}(N; \overline{p_a} + \eta_a), \\ N_{bn} &= \mathcal{B}(N; \overline{p_{bn}} + \eta_{bn}), \end{aligned} \tag{S1}$$

where η_a and η_{bn} are independent random Gaussian noises of mean 0 and of standard deviation σ_a and σ_{bn} , respectively. Note that here, for each citizen in a surrogate-municipality, probabilities to not vote and to put a null-blank vote are mutually independent.

Now, we can compare the average values $\overline{S}(N)$ of municipal involvement entropy in a sample of $\approx N$ surrogate-municipality-size, with $\overline{S}(N_{max})$ in the sample of the most populated municipalities. We find

that the difference is less than 1% when $N \gtrsim 2000$. In other words, for municipality-size greater than around 2000, statistical fluctuations due to finite size effects are negligible for what concern the present study.

To conclude, we have seen that, for French electoral data, finite size effects do not affect significantly the municipal involvement entropy (i.e. by less than a 2% deviation) for N greater than 2000. Note that 2000 is much less than the typical municipality size of the most populated municipalities, for which the common value $S \approx 1$ is frequently found. Lastly, the same analysis done for other countries for which electoral data are also available at the polling station scale (see the SI Section A) give the same results (see e.g. mean values of S over the 100 most populated municipalities, at the municipality scale in Tab. S1, compared to those at ballot box scale in Tab. S3).

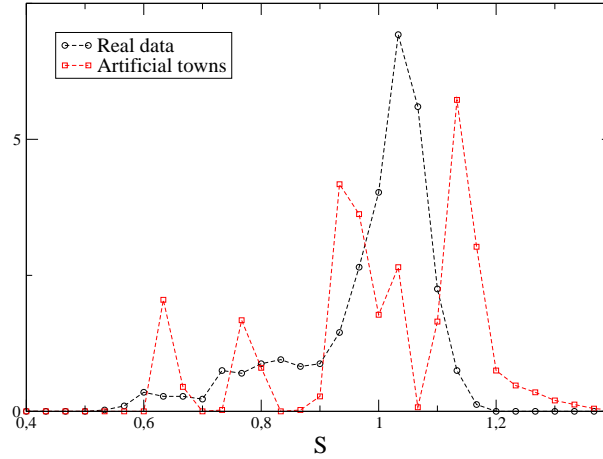


Figure S6. Histograms of S of the 100 most populated towns compared with 100 artificial towns (see text), in France, over elections since 2000.

Now, let us show that the shape of the distribution of S over the 100 most populated towns (which is sharply peaked near $S \approx 1$, apart from Austria) does not result from aggregating a large number of the citizen choices. In other words, the shape of the distribution of the involvement entropy for the 100 most populated towns (cf. Fig. 7-d) cannot be explained by a statistical bias due to a large number effect.

In order to see this point, 100 artificial town is created – in France, without the loss of generality. Each artificial town results from the aggregation over 300 real small municipalities of real numbers of registered voters (N), abstentionists (N_a), blank and null votes (N_{bn}) and votes according to the list of choices (N_c). In other words, an artificial town comes from the aggregation of real citizen choices who live in small municipalities. Each municipality is taken into account only once. These 100 French artificial towns have artificial aggregated registered voters (N) from 7000 to 330000, and is equal to 34000 in average. Fig. S6 allows one to compare the real distribution of S of the most populated French towns over elections since 2000 with the one which results from these 100 artificial towns. These two histograms are clearly different.

To conclude, the shape of the distribution of the involvement entropy of most populated towns (cf. Fig. 7-d) is not due to a bias rooted in aggregating a large number of citizen choices. The shape itself depends on real citizen choices who live in these towns.

D. Logarithmic three choices value, τ_3 , of polling stations

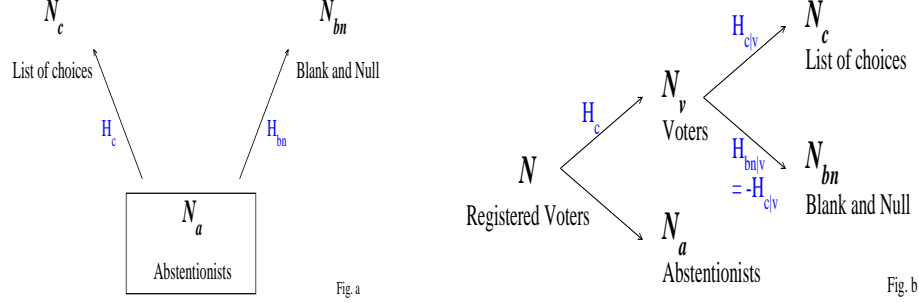


Figure S7

The aim of this section is to introduce a variable, called logarithmic three choices value, τ_3 , which takes into account the set of three ratios $\{p_a, p_c, p_{bn}\}$. First, distribution of τ_3 , over polling stations in the 100 most populated municipalities appears stable over time, and also similar between different countries. Secondly, we give arguments against two successive binary choice decisions leading to blank and null votes or votes according to the list of choices (i.e. to vote or not, then to cast a valid vote). This yields to consider together the three quantities p_a , p_c and p_{bn} in order to deal with the electoral involvement.

• Threshold decision rule of a unique binary choice.

Let an agent i , its decision ($n_i = 0$ or 1) is written as $n_i = \Theta(h_i + H)$, with $\Theta(x \geq 0) = 1$ and $\Theta(x < 0) = 0$. h_i is a kind of idiosyncratic term of the agent i . Idiosyncrasies are considered as uncorrelated between agents, and are described as independent random variables. H is a global field apply to agents, like a ‘Cultural field’ [23], etc. Note that there is no interaction between agents in this rough decision rule.

According to this decision rule, the aggregate value like the ratio p of the decision $+1$ over agents writes²³ as $p = \mathcal{P}_>(-H)$, where $\mathcal{P}_>(-H)$ means the cumulative distribution of idiosyncrasies h , i.e. $\mathcal{P}_>(-H) = \int_{-H}^{\infty} P(h)dh$. If idiosyncrasies are assumed to be distributed according to a logistic distribution, P , of zero mean and of unity width, it comes that

$$p = \mathcal{P}_>(-H) \xrightarrow{h: \text{logistic}} p = \frac{1}{1 + e^{-H}}, \text{ or, } H = \ln\left(\frac{p}{1-p}\right). \quad (\text{S2})$$

(Applied to voter turnout $p_v = 1 - p_a$, distribution of logarithmic turnout rates $\tau = \ln\left(\frac{p_v}{1-p_v}\right)$ across French municipalities is surprisingly stable over time [23], that allows to make predictions – confirmed by real measures [24, 25].)

• 2 mutually exclusive threshold decisions

Now, let us assume that the two decisions, (1) to vote according to the list of choices and (2) to cast a blank or null vote, are mutually exclusive. In other words, abstentionists are considered like a reservoir from which agents decide to make or not the choice (1) or the choice (2); moreover if they decide to do choice (1) (or conversely (2)), they do not decide anymore to make or not the choice (2) (or conversely (1)). Thus, let H_c the global field in favor of the choice (1) (to vote according to the list of choices), and p_c^0 the global ratio if choice (1) was unique, i.e. without any existence of choice (2) (see Fig. S7-a). Conversely, H_{bn} and p_{bn}^0 refer to choice (2) (to put a blank or null vote) if it was a unique choice. From

²³Finite size effects are here neglected.

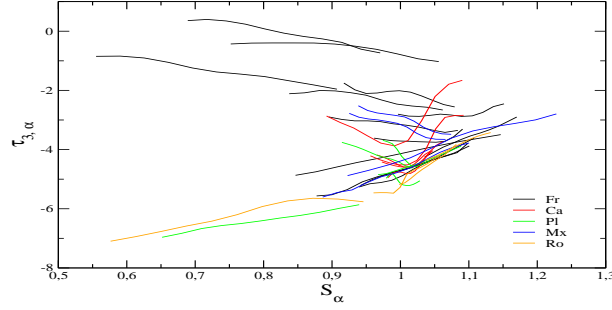


Figure S8. Logarithmic three choices value, τ_3 , with respect to involvement entropy S . Measures come from mean values of τ_3 and S over polling stations in each of the 100 most populated towns. Curves are smoothed. Note that there is not a one-to-one relation between τ_3 and S .

Eq. (S2), and still assuming a logistic distribution of idiosyncrasies, it comes

$$p_c^0 = \mathcal{P}_>(-H_c), \quad \text{thus, } H_c = \ln \left(\frac{p_c^0}{1 - p_c^0} \right),$$

$$p_{bn}^0 = \mathcal{P}_>(-H_{bn}), \quad \text{thus, } H_{bn} = \ln \left(\frac{p_{bn}^0}{1 - p_{bn}^0} \right). \quad (\text{S3})$$

In agreement with what precedes, choice (1) (or conversely (2)), is made over agents who have not decided to make the other choice (2) (or conversely (1)). By this way, observed ratios p_c and p_{bn} that exist when the two choice exist in the same time, are related to p_c^0 and p_{bn}^0 (that would exist if each one uniquely was existed) such that

$$p_c^0 = \frac{p_c}{1 - p_{bn}},$$

$$p_{bn}^0 = \frac{p_{bn}}{1 - p_c}. \quad (\text{S4})$$

Writing $\tau_3 = H_c + H_{bn}$, the sum of civic global fields applied to registered voters in this 3 choices process (leading to p_c , p_{bn} and p_a), Eqs. (S3,S4) yield to ²⁴

$$\tau_3 = \ln \left(\frac{p_c \cdot p_{bn}}{(p_a)^2} \right). \quad (\text{S5})$$

• τ_3 of polling stations in most populated towns

First, there is not a one-to-one relation between the logarithmic three choices value, τ_3 , and the involvement entropy S . Indeed, it is enough to invoke that the three ratios $\{p_a, p_c, p_{bn}\}$ play a symmetric role for S , and not for τ_3 . Fig. S8 plots τ_3 with respect to S for their average values over polling stations in each of the 100 most populated towns (see also Tab. S3 for basics statistics of S and τ_3 over polling stations in the 100 most populated municipalities).

Fig. S9 shows the pdf of the logarithmic three choices value τ_3 over different polling stations of the 100 most populated towns in each country (apart from Canadian ones because more than third of polling stations have $p_{bn} = 0$, which lead to their logarithmic three choices values τ_3 are undefined), i.e. the probability $P(\tau_3)d\tau_3$ that a given polling station, inside the 100 most populated towns, has τ_3 to within $d\tau_3$. Although the average $\langle \tau_3 \rangle$ over these polling stations varies quite substantially between elections (see Fig. S8 and Tab. S3), the shape of the distribution of $\tau_3 - \langle \tau_3 \rangle$ is quite constant for each country, and particularly for France and Mexico.

²⁴When one of the three ratios $\{p_a, p_c, p_{bn}\}$ is equal to zero, τ_3 is undefined.

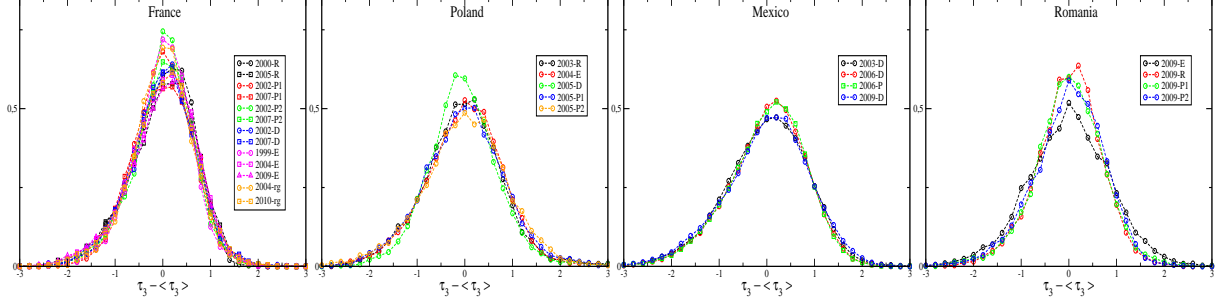


Figure S9. Distribution over polling stations of the 100 most populated towns of $P(\tau_3 - \langle \tau \rangle)$ for each election, where τ_3 is the logarithmic three choices value and $\langle \tau_3 \rangle$ its average value over all concerned polling stations.

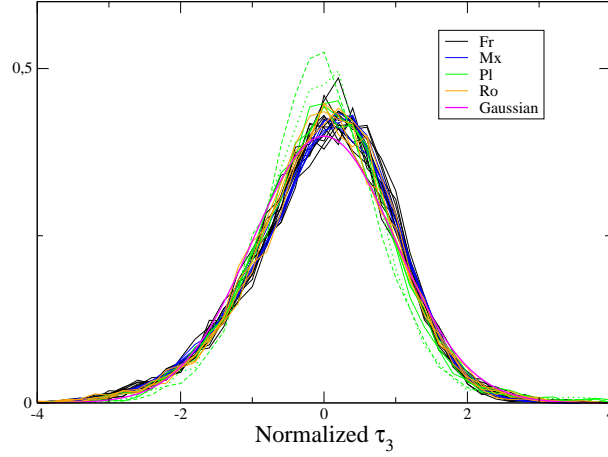


Figure S10. Distribution of normalized τ_3 over polling stations of the 100 most populated towns for 26 elections. The dotted line and the dashed line show respectively Pl-2003-R and Pl-2005-D elections. A normalized Gaussian is also plotted.

Moreover, distributions of normalized τ_3 (i.e. $\hat{\tau}_3 = \frac{\tau_3 - \langle \tau_3 \rangle}{\sigma}$, where $\langle \tau_3 \rangle$ and σ are respectively the mean value and the standard deviation of τ_3 over polling stations of the 100 most populated municipalities) appear to be similar with one another, as well in a same countries as in other countries, for French, Mexican, Romanian and half Polish elections (see Fig. S10). In fact, a Kolmogorov-Smirnov test where one only allows for a relative shift of the normalized distributions $P(\hat{\tau}_3)$, over polling stations of the 100 most populated towns, does not allow one to reject the hypothesis that the distribution $P(\hat{\tau}_3)$ is indeed the same for all elections (from France, Mexico, Romania and half ones in Poland).

Note that we have restricted this study to the polling stations of the 100 most populated towns because the common-value $S \approx 1$ of civic involvement appears for municipalities with high populations.

• Two successive binary threshold decisions

Now, let us show why it is preferable to consider two exclusive decisions (cf. Fig. S7-a) than two successive decisions (cf. Fig. S7-b) related to the civic-involvement of the electorate. For the latter case, the first binary decision is to vote or not to vote, and the second binary decision is to decide to cast a valid vote (according to the list of choices) or to put an invalid vote (i.e. a Blank or Null vote) knowing that the

considered agent is a voter (i.e. the first decision is to vote). In other words, the decision to put a vote according to the list of choices (or conversely to put a Blank or Null vote) results from two successive binary decisions: first to vote, second to put a valid vote (or conversely to put a Blank or Null vote) knowing that the first decision is to vote.

Let H_v the global field related to the first decision, i.e. to vote (see Fig. S7-b). Thus the ratio of voters, p_v , over registered voters writes as:

$$p_v = \mathcal{P}_>(-H_v), \quad \text{or,} \quad H_v = \ln\left(\frac{p_v}{1-p_v}\right). \quad (\text{S6})$$

(Remind that $p_v = 1 - p_a = p_c + p_{bn}$.) Let $H_{c|v}$ the global field related to the second binary decision, i.e. knowing that the agent is a voter, to put a vote according to the list of choices. So, the ratio of votes according to the list of choice over voters is written as

$$\frac{p_c}{p_v} = \mathcal{P}_>(-H_{c|v}), \quad \text{or,} \quad H_{c|v} = \ln\left(\frac{\frac{p_c}{p_v}}{1 - \frac{p_c}{p_v}}\right) = \ln\left(\frac{p_c}{p_{bn}}\right). \quad (\text{S7})$$

With the opposite convention to the previous one, the second decision to put a Blank or Null vote is such that $H_{bn|v} = -H_{c|v}$ (since $p_{bn}/p_v = \mathcal{P}_>(-H_{bn|v})$ and $H_{bn|v} = \ln\left(\frac{p_{bn}}{p_c}\right)$).

According to this two successive binary choices, the global field which leads a registered voter to put a Valid vote is $H'_c = H_v + H_{c|v} = \ln\left(\frac{p_v p_c}{p_a p_{bn}}\right)$; and to put a Blank or Null vote is $H'_{bn} = H_v + H_{bn|v} = \ln\left(\frac{p_v p_{bn}}{p_a p_c}\right)$. When Blank or Null ratio is very small ($p_{bn} \ll 1$), $p_c \simeq p_a$, hence leads to $H'_{bn} \simeq \tau_3$. So, statistics of H'_{bn} are very near to those of τ_3 .

If this successive binary decisions point of view is roughly correct, H'_{bn} and H'_c should share main features. Nevertheless, this is strongly rejected by real data. The shape of the distribution $H'_{bn} - \langle H'_{bn} \rangle$ is not constant at all (not shown in this paper) for each country over various elections over polling stations of the 100 most populated towns, and also confirmed by the Kolmogorov-Smirnov distance between two distributions. This allow us to prefer the hypothesis of two mutually exclusive binary decisions (which leads to τ_3 , and its surprising stability over time and countries) compared to the hypothesis of two binary decisions (which should lead to H'_c or H'_{bn} which should have the same features) about the civic-involvement.

E. Looking for signs of tension, through polling stations analysis

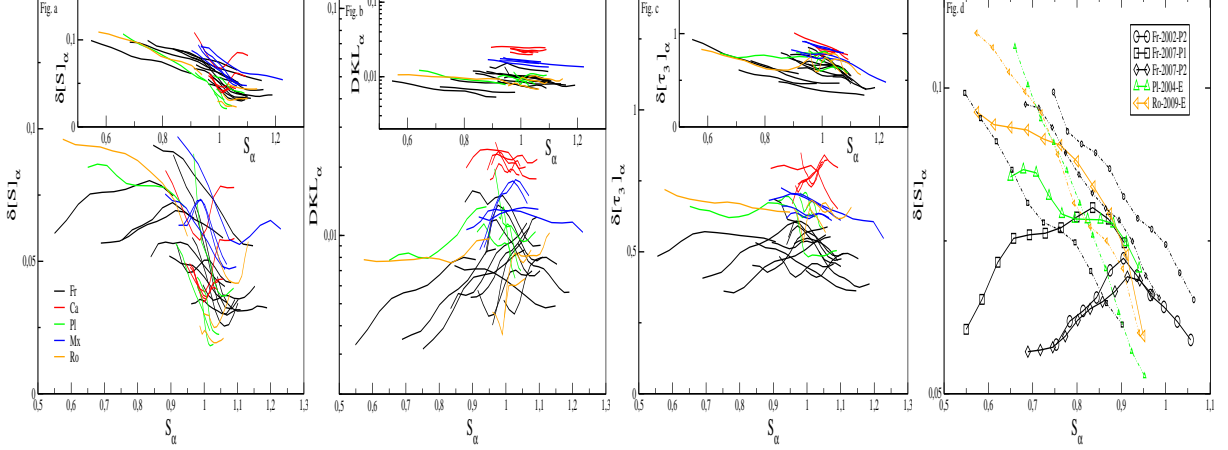


Figure S11. Civic-involvement heterogeneity (at the polling station scale) in a town with respect to the involvement entropy of the town. Curves are smoothed and concern the 100 most populated municipalities. Benchmark (see text) curves are plotted in the insets. Heterogeneity measures result from standard deviation of involvement entropy of polling stations (S11-a), KullbackLeibler divergence between a polling stations and other polling stations of the town (S11-b), and standard deviation of logarithmic 3 choices value of polling stations (S11-c). Fig. S11-d: same as Fig. S11-a, but restricted for the 5 elections which deviate the more from $S \approx 1$, where plain lines and dashed line plot respectively real data and benchmark curves.

This section seeks to detect some ‘tension’, in connection with the involvement entropy. We follow the assumption that ‘tension’ have some effects for polling stations heterogeneity inside a town. In other words, we try to detect some specific variation of polling stations heterogeneity in a given town, in connection with the involvement entropy of this town. Polling stations (inside a same town) heterogeneity is investigated by three different ways: (1) standard deviation of involvement entropies over all polling stations of the considered town; (2) KullbackLeibler divergence from one polling station compared to other polling station of the town; (3) standard deviation of the logarithmic three choices value (because the shape of its distribution is stable, see the SI Section D) over all polling stations of the town.

The analysis uses polling stations inside the 100 most populated towns (see the SI Section A for more details). Real results will be compared to a benchmark. The benchmark is based on the same heterogeneity of ratios p_a (idem for p_c , and p_{bn}) of polling stations for every town.

Let a town and a polling station of this town respectively called α and α_i . The polling station α_i has some measures, for instance its number of registered voters N_{α_i} , and the set of 3 ratios $\{p_{a,\alpha_i}, p_{c,\alpha_i}, p_{bn,\alpha_i}\}$ that provides its involvement entropy S_{α_i} and its logarithmic three choices value τ_{3,α_i} . The average over all the polling stations of the town (weighted by the number of registered voters, N_{α_i}), gives the corresponding value for the whole town α , e.g. the set of 3 ratios $\{p_{a,\alpha}, p_{c,\alpha}, p_{bn,\alpha}\}$, its involvement entropy S_α , and its logarithmic three choices value $\tau_{3,\alpha}$. The weighted (by the number of registered voters) standard deviation over all the polling stations α_i of the town α is written as $\delta[\cdot]_\alpha$, like for instance $\delta[p_a]_\alpha$, etc., $\delta[S]_\alpha$ and $\delta[\tau_3]_\alpha$. These quantify the heterogeneities within the town α .

Fig. S11-a plots involvement entropy heterogeneity of a town α (i.e. $\delta[S]_\alpha$) with respect to its involvement entropy (i.e. S_α). One should pay attention to the fact that $\delta[S]_\alpha$ going trough a minimum as $S_\alpha \approx 1$ could just be a consequence of $|dS|$ having a minimum near $p_{bn} \approx 0$ and $p_a \approx 0.5$. Hence the

benchmark presented here consists in comparing the empirical data with surrogate ones for which the heterogeneity in p_a , p_c and p_{bn} is the same for all municipalities, up to a binomial noise.

Here, the benchmark forces the same heterogeneity of the set of ratios $\{p_a, p_c, p_{bn}\}$ for every town, but keep their initial value of $\{p_a, p_c, p_{bn}\}$ for the whole town. In other words, let a town α , $\delta[p_a]_\alpha$, $\delta[p_c]_\alpha$ and $\delta[p_{bn}]_\alpha$ have the same values than in other towns; but $\{p_{a,\alpha}, p_{c,\alpha}, p_{bn,\alpha}\}$ are the real values of the town α , measured by the election.

The benchmark is realized as follows. First, we measure for each town α , $p_{a,\alpha}$ and $p_{bn,\alpha}$; and also $\delta[p_a]_\alpha$ and $\delta[p_{bn}]_\alpha$. The average values of heterogeneities $\delta[p_a]_\alpha$ and $\delta[p_{bn}]_\alpha$ over the 100 considered towns are respectively written as $\overline{\delta_{p_a}}$ and $\overline{\delta_{p_{bn}}}$. Secondly, we drawn from a binomial distribution, for each polling station α_i , its number of registered voters who do not take part to the election (N_{a,α_i}) and the number of Blank and Null votes (N_{bn,α_i}), such that:

$$\begin{aligned} N_{a,\alpha_i} &= \mathcal{B}(N_{\alpha_i} ; p_{a,\alpha} + \eta_a), \\ N_{bn,\alpha_i} &= \mathcal{B}(N_{\alpha_i} ; p_{bn,\alpha} + \eta_{bn}), \end{aligned} \quad (S8)$$

where η_a and η_{bn} are independent Gaussian noises of mean 0 and of standard deviation $\overline{\delta_{p_a}}$ and $\overline{\delta_{p_{bn}}}$ respectively, and N_{α_i} is the real number of registered voters of the polling station α_i of the considered town α . Note that we use a binomial distribution in order to take into account finite size effects; and here, for each citizen in a surrogate-polling station, probabilities to not vote and to put a null-blank vote are mutually independent.

Instead of making use of standard-errors, an alternative measure of heterogeneity is provided by making use of the so-called KullbackLeibler divergence which characterizes the difference between two probability distributions. For each polling station α_i of a given town α , we compute the divergence DKL_{α_i} from the polling station distribution P_{α_i} to the rest of the town, $Q_{\alpha_i} \equiv P_{\alpha-\alpha_i}$,

$$DKL_{\alpha_i} \equiv \sum_j P_{\alpha_i}(j) \log \frac{P_{\alpha_i}(j)}{Q_{\alpha_i}(j)} \quad (S9)$$

where, here and in the following, for any distribution we write $P(j)$, $j = 1, 2, 3$, instead of p_a, p_c, p_{bn} . Then we compute the mean KullbackLeibler divergence, DKL_α , of the town α by averaging over all polling stations, weighting by the corresponding number of registered voters, N_{α_i} .

$$DKL_\alpha \equiv \sum_i \frac{N_{\alpha_i}}{N_\alpha} DKL_{\alpha_i} \quad (S10)$$

This mean KullbackLeibler divergence DKL_α gives us a measure of heterogeneity of polling stations into a town α .

Fig. S11 compares benchmarks curves and empirical data. It appears that, the smaller the involvement entropy S (with $S \lesssim 0.85$), the smaller the involvement entropy heterogeneity at the polling station level (see more specifically Fig. S11-d). Heterogeneity of polling stations in a same town is measured via three different ways: standard deviations of the involvement entropy and the logarithmic three choices ratio, and also via the KullbackLeibler divergence. In other words, the more the town is “ordered” (for its electorate civic-involvement), the more the town is homogeneous (at the polling station scale, and still for a civic involvement point of view). Note also that this point is particularly clear when the ratio p_c is high (e.g. for 3 French elections), compared to cases where p_a are high (e.g. for European Parliament elections in Romania and Poland). It can also be noted that often heterogeneity of involvement entropies of polling stations inside towns ($\delta[S]_\alpha$) has a significantly minimal value when their involvement entropies (S_α) are around 1, and this minimization is much more marked for real data than for benchmark ones.

F. Disentangling Blank votes from Null votes

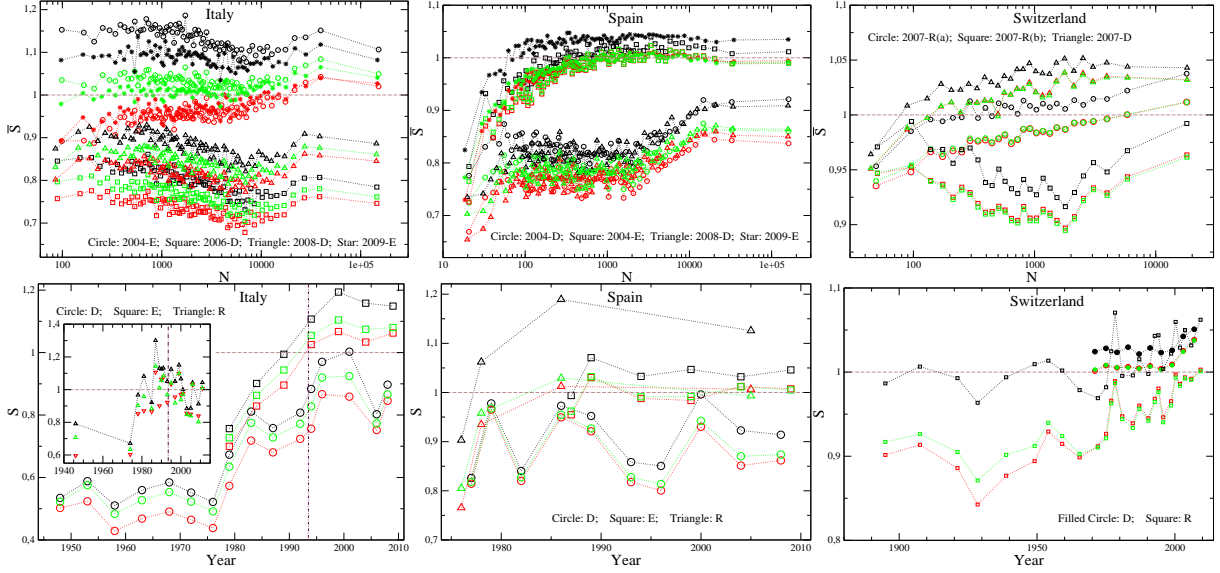


Figure S12. Blank votes are grouped with: (1) Null votes (like in the main text, cf. Eq. (2)) in black; (2) Valid Votes or another vote included in the list of choices (cf. Eq. (S12)) in red; (3) citizens who do not take part to the election (cf. Eq. (S13)) in green. Top: mean values of S , $S_{b=c}$, $S_{b=a}$, over bins with around 100 municipalities of size $\approx N$ (like in Fig. 4). Bottom: Evolution in time of S , $S_{b=c}$, $S_{b=a}$ (with the same scale of aggregate data as in Fig. 5). For the sake of clarity, standard deviations over Swiss *Cantons* and Spanish *Comunidades autónomas* are not shown. Each point (R) for Swiss graph gives the average of around 20 Swiss referendums. The end of Italian compulsory voting is shown by a vertical line.

Italy²⁵, Spain and Switzerland²⁶ are countries for which Blank votes are distinguished from Null votes. Let N_b and N_n the number of citizens who respectively vote Blank and Null amongst N registered voters (of one municipality, *Canton*, *Comunidad autónoma*, the whole country). Ratios, or probabilities, to respectively vote Blank and Null are

$$p_b = \frac{N_b}{N}, \quad p_n = \frac{N_n}{N}. \quad (\text{S11})$$

In such cases, it is legitimate to consider that Blank votes should be categorized with votes in favor of one of the proposed choices to the election. Then, the Blank vote has not a ‘marginal’ involvement meaning, like previously, but its citizen involvement is similar to another Valid vote according to the list of choices of the election. One should then consider a modified involvement entropy, defined from the 3-set ratios (of sum unity) $\{p_a, (p_c + p_b), p_n\}$, that is

$$S_{b=c} = -p_a \log(p_a) - (p_c + p_b) \log(p_c + p_b) - p_n \log(p_n). \quad (\text{S12})$$

Alternatively, one may consider that Blank votes lose their ‘marginal’ aspect in citizen involvement, and should be categorized as votes from citizen who do not take part to the election. Then the relevant

²⁵We only analyze the first question asked in a Referendum. Senate elections are not shown in Fig. S12-below because they are very similar to Chamber of Deputies (D) elections.

²⁶Chamber of deputies elections (D) distinguish, in our database, Blank vote between Null votes since 1971; and since 1887 for *votations* (or referendums).

modified involvement entropy, defined from the 3-set ratios (still of sum unity) $\{(p_a + p_b), p_c, p_n\}$, writes as

$$S_{b\equiv a} = -(p_a + p_b) \log(p_a + p_b) - p_c \log(p_c) - p_n \log(p_n). \quad (\text{S13})$$

Figures S12 shows for Italy, Spain and Switzerland, the involvement entropy, S , and the modified versions, $S_{b\equiv c}$ and $S_{b\equiv a}$: (1) for municipalities and with respect to the municipality-size N (as in Fig. 4); (2) for the whole country (directly for Italy, and as a weighted mean by population-size over 25 or 26 Swiss *Cantons* and 17 or 19 Spanish *Comunidades autónomas*) as a function of time (as in Fig. 5). Fig. 8 shows the modified involvement entropy $S_{b\equiv c}$ ($S_{b\equiv a}$ which is not shown, is very close to $S_{b\equiv c}$), with respect to the involvement entropy S , for ~ 530 Swiss Referendums.

Figures S12 and 8 exhibit some trends and regularities that depend on the values of involvement entropy S . (1) When $S < 1$ (e.g. in Italian and Spanish Chamber of Deputies elections, both at municipality scale or at large scale of aggregate data), modified involvement entropies are smaller than S . This means a greater order of the modified citizen involvement. It can be interpreted as follows: the loss of nuance or specificity (for citizen involvement) that Blank vote have, implies a greater polarization or heterogeneity of the electorate, still split into 3 groups. (2) When $S \approx 1$, two different cases arise. First, for Spanish European Parliament elections, Swiss Chamber of Deputies elections and Referendums (uniquely for the latter, since the 2000s), both at municipality scale or at large scale of aggregate data: the modified involvement entropies are slightly lower than S , but still ≈ 1 . Second, for earlier Swiss Referendums, and particularly before the 1960s: $S_{b\equiv c}$ (or $S_{b\equiv a}$) are lower than S , but not slightly lower. (3) When $S > 1$ and $S \not\approx 1$ (e.g. for Italian European Parliament elections, both at municipality scale or at large scale of aggregate data, and Spanish Referendums, particularly 1986 and 2005 ones, at provincial scale), modified involvement entropies are still lower than S . But one more time, it is surprising to notice that modified involvement entropies are such that $S_{b\equiv c} \approx 1$ (or $S_{b\equiv a} \approx 1$). It can be explained as follows: subtlety or specificity of citizen involvement due to Blank votes means an increasing of disorder of the electorate involvement. The loss of this subtlety or specificity (i.e. considering Blank votes like another vote in favor of the list of choices, or like another abstentionist) implies a loss of ‘tension’ contained in electoral campaign. And strikingly, this loss of ‘tension’ provides a new entropy, like the usual common-value of involvement entropy, ≈ 1 .

Note that above items (1) and (3) (i.e. when significantly $S < 1$ or $S > 1$), pointed out in Fig. S12, are clearly shown in Fig. 8 for Swiss Referendums. (In the Fig. 8, S' means $S_{b\equiv c}$, which is very near to $S_{b\equiv a}$ on average.) Note also that the surprising plateau (which provides modified involvement entropies equal to ≈ 1 , on average, when $S \gtrsim 1.05$) does not exist, in our database, for most populated municipalities. For the latter case (not shown), the around 100 most populated municipalities for which $S \gtrsim 1.05$, uniquely provides $S_{b\equiv c}$ (or $S_{b\equiv a}$) lower than S such that, on average, $\overline{S_{b\equiv c}} \approx 1$ (or $\overline{S_{b\equiv a}} \approx 1$), but without a plateau.

Lastly, this study does not allow us to know whether it is more meaningful (according to the entropy of the electorate involvement) to consider Blank vote like another vote proposed in the list of choices or like another abstentionist vote. Nevertheless, in our database, Blank votes seem more meaningful than Null Votes in Spain and in Switzerland. Indeed, when p_n and p_b are interchanged between each other in Eqs. (S12) or (S13), above item (3) (when significantly $S > 1$, then the modified involvement entropy is ≈ 1) does clearly not exist for Spanish and Swiss Referendums.

To conclude, let us recall the main point of this section: when involvement entropy S does not obey to the common occurrence $S \approx 1$ for high population-size municipalities, or at large aggregate scale, because the citizen involvement of the electorate is too much disordered (i.e. significantly $S > 1$), then the modified involvement entropy (by the loss of the specificity of Blank votes) takes on average the same common value ≈ 1 .

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